



DEPARTMENT OF
COMPUTER SCIENCE

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BSc in Computer Science and Engineering

MICROBABEL

A MULTI-PROTOCOL APPROACH FOR RESILIENT
AND DECENTRALIZED IOT NETWORKS

MASTER IN COMPUTER SCIENCE AND ENGINEERING
SPECIALIZATION IN DISTRIBUTED AND PARALLEL SYSTEMS

NOVA University Lisbon

Draft: January 30, 2026



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ABSTRACT

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1. What is the problem?
2. Why is this problem interesting/challenging?
3. What is the proposed approach/solution/contribution?
4. What results (implications/consequences) from the solution?

Keywords: One keyword · Another keyword · Yet another keyword · One keyword more
· The last keyword

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1. Qual é o problema?
2. Porque é que é um problema interessante/desafiante?
3. Qual é a proposta de abordagem/solução?
4. Quais são as consequências/resultados da solução proposta?

Palavras-chave: Primeira palavra-chave · Outra palavra-chave · Mais uma palavra-chave · A última palavra-chave

CONTENTS

Glossary	vii
Acronyms	viii
1 Introduction	1
2 Related Work	6
2.1 Wireless Communication Technologies	8
2.1.1 Application-Layer Protocols	8
2.1.2 Physical- and Link-Layer Technologies	8
2.1.3 Discussion	9
2.2 Peer-to-peer (P2P) Mesh Networking and Topology Management	10
2.2.1 Peer Discovery and Topology Maintenance	10
2.2.2 Routing in Partitioned and Intermittently Connected Networks	12
2.2.3 Limitations of Centralized Coordination	12
2.2.4 Discussion	13
2.3 Multi-Protocol Communication and Adaptive Protocol Selection	14
2.3.1 Multi-Channel Resilience Architectures	14
2.3.2 Protocol Heterogeneity on Embedded Platforms	15
2.3.3 Runtime Adaptive Protocol Selection	16
2.3.4 Discussion	16
2.4 Decentralized Synchronization and Time Coordination	16
2.4.1 Master-Based Synchronization as Counterexample	17
2.4.2 Gossip-Based Masterless Synchronization	17
2.4.3 Coordination Through Passive View Maintenance	18
2.4.4 Discussion	18
2.5 Lightweight Data Compression and Energy Efficiency	19
2.5.1 Prediction-Based Data Reduction	19
2.5.2 Lightweight Compression	20
2.5.3 Receiver-Side Energy Considerations	22

2.5.4	Discussion	22
2.6	Security Considerations and Scope Delimitation	23
2.7	Summary	24
3	Solution Architecture	26
3.1	Recap	26
3.1.1	The Problem	26
3.1.2	Technical Challenges	26
3.2	System Requirements	27
3.2.1	Overview of μ - Babel 's Approach	27
3.3	Operating Conditions and Failure Modes	28
3.3.1	Target Operating Environments	29
3.3.2	Expected Failure Modes	29
3.4	Hardware Platform	30
3.4.1	Class-1: Battery-Powered Sensor Nodes	30
3.4.2	Class-2: Optional Aggregation Nodes	32
3.5	Software Stack and Development Environment	32
3.5.1	Operating System Layer	32
3.5.2	Platform SDKs and Hardware Abstractions	33
3.5.3	Concurrency Model	33
3.6	System Model	34
3.6.1	Core Infrastructure Layer	34
3.6.2	Discovery & Topology Layer	35
3.6.3	Coordination Layer	36
3.6.4	Data Management Layer	38
3.6.5	Component Interactions	39
4	Progress Report and Planning	40
4.1	Progress Report	40
4.1.1	Use Cases	40
4.2	Planning	40
4.2.1	Validation	40
4.2.2	Task Scheduling	40
	Bibliography	43

LIST OF FIGURES

4.1	GANTT	42
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LIST OF TABLES

2.1	Protocol Characteristics	9
2.2	Related work summary	24
3.1	Class-1 Device Specifications	30
3.2	Supported Communication Protocols	31
3.3	Class-2 Device Specifications	32

GLOSSARY

ESP-NOW	Wireless communication protocol developed by Espressif. Enables direct and low-power communication between ESP-family devices without the need for a router. (<i>pp. 4, 9, 15, 31, 35, 37</i>)
LoRa	From "long range", proprietary radio communication technology (<i>pp. 3, 4, 9, 13, 14, 19, 31, 35, 37, 38</i>)
Modbus	An application-layer client/server serial communication protocol, widely used in industrial automation systems. (<i>p. 15</i>)
MQTT	Lightweight publish/subscribe messaging protocol designed for constrained devices and low-bandwidth networks. Previously stood for "Message Queuing Telemetry Transport", but since 2013 it does not stand for anything in particular (<i>pp. 8, 16, 22, 28, 32</i>)
Thread	Low-power, mesh networking technology based on the IEEE 802.15.4 standard. Increasingly adopted for smart-home applications. (<i>p. 31</i>)
μ-Babel	MicroBabel (<i>pp. iv, 2–4, 7–9, 13, 16, 19, 22–25, 27–34</i>)
ZigBee	Low-power, low-data rate wireless communication specification based on the IEEE 802.15.4 standard. Designed for short-range communication in mesh, star or tree network topologies. (<i>pp. 4, 9, 15, 31, 35</i>)

ACRONYMS

AES	Advanced Encryption Standard (<i>pp.</i> 15 , 17 , 23)
API	Application programming interface (<i>p.</i> 33)
ARIMA	Autoregressive integrated moving average (<i>pp.</i> 20 , 22)
AWCT	Always Connected Things (<i>p.</i> 14)
BLE	Bluetooth Low Energy (<i>pp.</i> 3 , 4 , 9 , 10 , 13–15 , 19 , 31 , 33 , 35 , 37 , 38)
CA	Certificate authority (<i>p.</i> 7)
CoAP	Constrained Application Protocol (<i>pp.</i> 8 , 22 , 28 , 32)
CPU	Central processing unit (<i>pp.</i> 30 , 32)
CR	Compression ratio (<i>p.</i> 21)
DNS	Domain Name System (<i>p.</i> 1)
DODAG	Destination-oriented directed acyclic graph (<i>p.</i> 11)
DTN	Delay-tolerant networking (<i>pp.</i> 11 , 12 , 14)
ESP-IDF	Espressif IoT Development Framework (<i>p.</i> 33)
FHSS	Frequency-hopping spread spectrum (<i>pp.</i> 15 , 17)
FIRE	Fast Integer REgression (<i>p.</i> 21)
GASE	Group Authentication Scheme at the Edge (<i>p.</i> 23)
GPIO	General-purpose input/output (<i>p.</i> 33)
GPS	Global Positioning System (<i>p.</i> 19)
HAL	Hardware Abstraction Layer (<i>p.</i> 33)
HF	High frequency (<i>p.</i> 14)
HTTP	Hypertext Transfer Protocol (<i>pp.</i> 12 , 22)
HVAC	Heating, ventilation, and air conditioning (<i>p.</i> 6)

I²C	Inter-Integrated Circuit (<i>p. 33</i>)
IoT	Internet of Things (<i>pp. 1–3, 6–12, 14, 19, 22–24, 26, 28, 30, 31</i>)
KDC	Key distribution center (<i>p. 7</i>)
LE	Low Energy (<i>p. 31</i>)
LoRaWAN	LoRa Wide Area Network (<i>pp. 10, 14, 16</i>)
LPKM	Lightweight Polynomial-based Key Management (<i>p. 23</i>)
LPWAN	Low-Power Wide Area Network (<i>p. 14</i>)
MAC	Message Authentication Code (<i>p. 23</i>)
MDL	Minimum description length (<i>pp. 22, 23</i>)
novathesis	NOVathesis L ^A T _E X (<i>pp. i, ii</i>)
NTP	Network Time Protocol (<i>p. 19</i>)
NVIS	Near Vertical Incidence Skywave (<i>pp. 14–16</i>)
OS	Operating system (<i>pp. 32, 33</i>)
OSPF	Open Shortest Path First (<i>p. 28</i>)
P2P	Peer-to-peer (<i>pp. iii, 3, 4, 7, 10–13, 15, 27, 29–31</i>)
PKI	Public key infrastructure (<i>p. 24</i>)
PLR	Packet loss ratio (<i>p. 11</i>)
PRNG	Pseudorandom number generator (<i>pp. 15, 17</i>)
PSRAM	Pseudostatic random-access memory (<i>p. 30</i>)
PUF	Physically Unclonable Function (<i>pp. 23, 27</i>)
QoS	Quality of Service (<i>pp. 4, 7, 37</i>)
RAM	Random-access memory (<i>p. 32</i>)
RF	Radio frequency (<i>p. 14</i>)
RGCS	Randomized gossip-consensus-based sync (<i>pp. 17–19, 25, 37</i>)
RLE	Run-length encoding (<i>pp. 21–23, 38</i>)
RPL	Routing Protocol for Low-Power and Lossy Networks (<i>pp. 11–14</i>)
RSSI	Received signal strength indicator (<i>pp. 10, 15</i>)
RTOS	Real-time operating system (<i>p. 32</i>)
SBC	Single-board computer (<i>p. 32</i>)
SDK	Software Development Kit (<i>pp. 32, 33</i>)

SDN	Software-defined networking (<i>pp. 7, 16, 24</i>)
SF	Spreading Factor (<i>p. 9</i>)
SPI	Serial Peripheral Interface (<i>pp. 31, 33</i>)
SRAM	Static random-access memory (<i>p. 30</i>)
SSID	Service Set Identifier (<i>p. 9</i>)
TCP	Transmission Control Protocol (<i>pp. 7, 11, 13, 25</i>)
UART	Universal asynchronous receiver-transmitter (<i>p. 33</i>)
UDP	User Datagram Protocol (<i>p. 8</i>)
UUID	Universally unique identifier (<i>p. 35</i>)
WSN	Wireless Sensor Network (<i>pp. 11, 14, 15, 17</i>)

INTRODUCTION

The proliferation of Internet of Things (IoT) devices has changed how we monitor and interact with physical spaces – from smart homes to industrial facilities – with the global IoT market reaching 18.5 billion connected devices in 2024, and projected to reach 39 billion by 2030 [41], with industry analysts estimating that IoT technologies could unlock between \$5.5 and \$12.6 trillion in economic value by the same year [10].

However, current IoT systems remain fundamentally dependent on continuous Internet connectivity and centralized cloud infrastructures [52, 21]. The dominant architectural pattern across the industry relies on edge devices collecting data and transmitting it to remote cloud servers for processing, storage, and execution of other control logic. This approach delivers scalability and ease of management but creates a fundamental dependency on network availability.

This centralized model introduces several concerns beyond just connectivity: operational costs for cloud services create economic dependencies on third-party providers and often lead to vendor lock-in, data sovereignty issues arise when sensitive information must travel to and reside on external infrastructure, and system resilience becomes fundamentally tied to the availability of these remote services.

While cloud platforms can benefit IoT deployments by providing additional computational resources and storage capacity, systems that *depend* on constant cloud connectivity sacrifice local autonomy and introduce single points of failure. This cloud-centric model has enabled rapid IoT adoption, but introduces critical vulnerabilities across scenarios where continuous connectivity cannot be guaranteed.

Recent infrastructure failures illustrate the fragility of cloud-dependent architectures: the AWS US-EAST-1 outage in October 2025 [44] disrupted services from banking to smart homes worldwide, demonstrating how a Domain Name System (DNS) configuration error in a *single region* can have cascading effects, leading to global disruptions [42].

Attempts have been made to address these challenges through incremental improvements, such as shifting focus to edge computing in order to reduce latency [21, 52, 37], redundant cloud regions for availability [29, 28], and hybrid architectures that combine local and remote processing [1, 12, 24].

However, these solutions remain fundamentally tied to the assumption of eventual connectivity, and often increase system complexity without eliminating the core dependency. This leads to shortcomings in key areas that demand more fundamental architectural reconsideration:

Limited suitability for hazardous environments:

Remote or dangerous locations (industrial sites, disaster-prone areas) require systems that can operate reliably without constant human intervention or stable network infrastructure;

Lack of autonomous operation:

Device deployment and operation en masse can be brittle, with little tolerance for individual node failures in the IoT infrastructure, which is naturally susceptible to network failures and intermittent connectivity;

Privacy and data sovereignty concerns:

Cloud platforms and other third-parties are oftentimes an unavoidable middle layer between end devices and end users, raising questions about data processing and control.

While these challenges affect both IoT and domotics systems alike – from industrial monitoring to residential automation – they become particularly acute in disaster response and emergency scenarios, where communication infrastructure might become completely unavailable precisely when needed most, rapid deployment with minimal configuration becomes essential, and autonomous operation transitions from desirable to indispensable.

During earthquakes, floods, and natural or human-driven disasters, the need for real-time sensor data (structural integrity, air quality, evacuation routes) and bidirectional communication (threat alerts, user feedback) becomes critical, yet traditional infrastructure often fails first, eliminating both cloud connectivity and local network access points that deployed devices rely on, making support systems for these scenarios unavailable or non-operational.

A pragmatic issue in the context of smart homes is that without cloud connectivity, users cannot interact with their appliances, such that during a Wi-Fi outage smart lights become uncontrollable despite all hardware being physically present. While merely inconvenient domestically, this architectural dependency has critical implications in other contexts and μ -**Babel** aims at addressing such scenarios across different application domains: from residential systems requiring local control, to building monitoring infrastructures that must operate during emergencies, to disaster response networks where autonomous operation becomes essential when traditional infrastructure fails.

While cloud platforms like Amazon Alexa, Google Home, and Microsoft Azure IoT Hub offer convenience for data processing and remote device control, their inherent dependence on continuous Internet connectivity introduces some critical limitations:

increased latency from round-trip communication to distant servers [37, 33], reduced availability during network disruptions or cloud outages [49, 28], and poor fault-tolerance when infrastructure fails [3].

These characteristics make cloud-centric architectures fundamentally unsuitable for scenarios that require or prioritize local operation, autonomous behavior, or guaranteed responsiveness during emergencies.

The Babel Ecosystem [18] partially addresses these limitations by enabling devices to operate autonomously without cloud infrastructure, while still supporting cloud integration when connectivity is available and desired.

In this document we propose μ -Babel, a lightweight framework targeting embedded platforms (e.g., ESP32, Raspberry Pi Pico) aimed at developing resilient, multi-protocol, and decentralized IoT systems that can operate autonomously during infrastructure failures.

μ -Babel will integrate with the broader Babel Ecosystem, which runs on hardware with greater resources such as full Raspberry Pi boards, desktops/laptops, or servers. This enables a heterogeneous architecture where resource-constrained edge devices can seamlessly interoperate with computational nodes for additional data aggregation, processing, and large-scale coordination.

FIXED: "class-based"

Note on terminology: Throughout this work the term *multi-protocol* is used to refer to heterogeneity in communication technologies (i.e., Bluetooth Low Energy (BLE), LoRa, Wi-Fi, etc.) that have distinct link/physical layers, while *multi-channel* refers to frequency diversity within a single protocol (e.g., 2.4GHz Wi-Fi channels 1-13, LoRa frequency hopping). Multi-protocol operation provides technology diversity for resilience; multi-channel operation provides frequency diversity for both anti-jamming and additional throughput.

Main Research Questions

In this work, we focus on three main research questions:

How can these systems maintain communication and full operation when traditional infrastructure fails?

Conventional deployments of networked devices often heavily rely on Wi-Fi access points, cellular towers, or other centralized infrastructure that can easily become unavailable during disasters, or is altogether unreliable in remote locations. μ -Babel will address this by leveraging a multi-protocol communication approach, supporting a diverse set of wireless protocols that can operate independently of infrastructure.

By enabling adaptive protocol selection and Peer-to-peer (P2P) mesh formation, devices can establish alternative communication paths when primary channels fail, allowing them to sustain information exchange and system availability.

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How can device heterogeneity be leveraged to create and orchestrate these systems?

Deployments naturally comprise devices with varying capabilities, from simple resource-constrained nodes, to more capable gateways that might act as aggregators or bridges. Rather than treating this heterogeneity as a limitation, μ -Babel will exploit it through capability-aware protocols that allow devices to negotiate roles dynamically via autonomous and automatic discovery.

Resource-rich nodes can serve as data aggregation and processing points, or bridges between a remote deployment and traditional network infrastructure (if desired), while simpler devices focus on sensing and actuation, creating a resilient multi-tier architecture.

How can we achieve (near-)zero-configuration deployment in diverse environments?

Straightforward deployments are essential where users cannot perform complex configurations – from smart homes and factories to hazardous areas. μ -Babel will provide automatic peer discovery across multiple protocols, self-organizing network formation, and decentralized coordination mechanisms that eliminate the need for pre-configured coordinator nodes or manual network planning.

Devices will autonomously establish connectivity with each other, negotiate protocols, and begin operation upon being activated, with minimal configuration effort to enable rapid deployment even in hard-to-reach locations.

Expected Contributions

We plan to make the following main contributions:

- A decentralized architecture supporting multiple communication protocols (Wi-Fi, BLE, LoRa, ZigBee, ESP-NOW) with adaptive switching based on Quality of Service (QoS) requirements, resource availability and device capabilities;
- A resource-efficient programming framework for embedded platforms that enables autonomous operation without central coordination, providing abstractions for multi-protocol communication, peer discovery, and opportunistic data forwarding;
- A proof-of-concept implementation demonstrating infrastructure-independent operation and automatic disaster-mode failover in a real-world deployment.

Document Structure (Roadmap)

The remainder of this document is arranged as follows:

Chapter 2 Reviews widely-used communication protocols and related work in P2P networking, multi-protocol communication, decentralized synchronization, data compression, and security.

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Chapter 3 Presents the system requirements, operating conditions, failure modes, hardware and software platforms, and the architectural model of the proposed solution.

Chapter 4 Summarizes the current implementation status and outlines the development and evaluation plan.

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RELATED WORK

Internet of Things (IoT) and home automation (domotics) systems share the same technological building blocks (wireless sensors, embedded devices, network communication) but diverge significantly in their operational focus and architectures:

FIXED: added citations

IoT systems [14] typically focus on **data collection and monitoring**, by streaming sensor readings to centralized platforms for analysis and processing, with control functions often being a secondary concern.

Domotics systems [19, 16] instead prioritize **real-time control and actuation** over physical spaces, where responsiveness and local autonomy are paramount for end user experience and privacy.

This distinction gains additional relevance when focusing on resilience requirements: while IoT deployments may tolerate delayed data aggregation and/or temporary connectivity issues in monitoring scenarios, domotics applications (such as emergency lighting control or Heating, ventilation, and air conditioning (HVAC) management) demand immediate local response regardless of network conditions.

Both domains, however, suffer from a common vulnerability when confronted with infrastructure failures during disasters: their predominantly cloud-centric architectures collapse precisely when autonomous operation becomes indispensable.

TOFIX: em vez de "iot systems": smart systems?

The challenges faced by disaster-resilient IoT systems span multiple dimensions:

Infrastructure failures eliminate access points that devices depend on for coordination and operation;

Intermittent connectivity creates network partitions where subgroups of devices must operate autonomously without global state synchronization;

Resource constraints limit the computational, memory and energy budgets available for implementing sophisticated resilience mechanisms in embedded platforms.

These challenges are fundamentally architectural: existing IoT and domotics systems are designed around the assumption of stable infrastructure, treating network partitions and coordinator failures as transient anomalies rather than the norm.

Cloud-centric architectures place control decisions in remote servers, creating dependencies that cannot be satisfied when connectivity fails. Coordinator-based topologies – whether using dedicated gateways, master nodes, Software-defined networking (SDN) controllers, among others – concentrate the risk of failure in single points that can paralyze entire systems if compromised or disconnected. Even ostensibly distributed systems often rely on persistent connections (i.e., Transmission Control Protocol (TCP)) over a single medium, or heavy middleware platforms that exceed the capabilities of embedded devices and, once more, assume some form of infrastructure availability.

The fundamental point of contention is between resilience requirements (autonomy during infrastructure collapse) and resource constraints (limited computation, memory, and energy). Traditional approaches address one at the cost of the other: cloud platforms provide sophisticated coordination but fail during outages; purely local systems avoid external dependencies but struggle with inter-device coordination and protocol heterogeneity with low resource usage.

Disaster-resilient systems require architectural choices that prioritize autonomous Peer-to-peer (P2P) connectivity as the baseline mode of operation rather than having it as a secondary – even exceptional – fallback mechanism.

Naturally, this requires a fundamental rethinking of IoT systems design: how devices discover and communicate with each other using diverse protocol stacks (without depending on centralized control), how they achieve temporal synchronization and coordination without master beacons, how to optimize the usage of limited available resources, and how to maintain secure operation throughout without persistent access to inherently centralized components such as Certificate authorities (CAs) or Key distribution centers (KDCs).

TOFIX: general statement.. não encontro citações to back this up exactly, sinto que é kind of a given

The following sections briefly cover related work across five areas that collectively enable autonomous operation: (Section 2.1) common wireless communication technologies across application, link, and physical layers; (Section 2.2) P2P mesh networking and topology management for autonomous network formation without coordinator dependencies; (Section 2.3) multi-protocol communication and adaptive selection based on Quality of Service (QoS) and device capabilities; (Section 2.4) decentralized synchronization mechanisms for coordinating multi-protocol communication without master nodes; (Section 2.5) lightweight data compression and energy optimization techniques to extend autonomous operation duration and device lifetime.

For each area, we examine how existing approaches handle (or fail to handle) infrastructure failure scenarios, identify architectural assumptions that conflict with disaster-resilient requirements, and position μ -Babel's contributions relative to the current state of the art.

Additionally, security and privacy considerations are presented in Section 2.6, which

FIXED: separei section sobre security, no longer part of "autonomous operation" areas per se

outlines the scope of this work in regard to higher-layer safety mechanisms.

2.1 Wireless Communication Technologies

A core pillar of resilient and decentralized IoT systems is the heterogeneity of available communication capabilities and wireless protocols across different layers of the communication stack. Different technologies naturally expose different trade-offs in terms of infrastructure requirements, range, throughput, discovery mechanisms, and energy consumption – all of which directly influence the design choices around their usage.

There is no single communication protocol that fully addresses the requirements of infrastructure-free and disaster-resilient operation. As such, it is important to establish the capabilities and limitations of each one, as well as their potential for harmonious cooperation within a unified system.

2.1.1 Application-Layer Protocols

At the application layer, two messaging protocols stand out given their wide-spread adoption in IoT deployments:

MQTT: A lightweight publish/subscribe messaging protocol designed for constrained devices and unreliable, low-bandwidth networks. It relies on a centralized *broker* to mediate communication between publishers and subscribers – essential, a server.

Constrained Application Protocol (CoAP) [8, 39]: A RESTful, request/response protocol designed for constrained environments. Operates over User Datagram Protocol (UDP) and assumes the presence of well-known endpoints or proxy servers for resource discovery and access.

Both protocols inherently assume the availability of stable endpoints, and rely on continuous network connectivity in order to operate. These assumptions conflict with the infrastructure-less and failure-prone environments targeted by μ -Babel. For this reason, while such protocols are of interest for broader interoperability (see Non-Functional Requirements in Section 3.2.1), they are not a part of the foundational protocols employed by this work.

2.1.2 Physical- and Link-Layer Technologies

This section reviews the physical- and link-layer wireless technologies relevant to resilient and decentralized IoT deployments. While the previous section addressed higher-layer messaging protocols, the technologies discussed here provide the foundational communication stack that determines the behavior of μ -Babel.

Table 2.1 details the basic operational characteristics of the wireless communication technologies considered in this work.

FIXED: moved protocols para aqui + MQTT and CoAP overview

get actual power draw values from somewhere

Table 2.1: Protocol Characteristics

Protocol	Discovery	Range ¹	Frequency	Data Rate	Power Draw
Wi-Fi	SSID scan	50-100 m	2.4 GHz	1-50 Mbps	High
BLE	Advertising	10-50 m	2.4 GHz	1-2 Mbps	Low
ESP-NOW	Peer list	50-100 m	2.4 GHz	1 Mbps	Very Low
ZigBee	Beacon	10-100 m	2.4 GHz	250 Kbps	Very Low
LoRa	Preamble detection	2-10 km	433/868/915 MHz ²	0.3-50 Kbps ³	Low ³

¹ Outdoor line-of-sight; indoor ranges typically 30-50% lower

² Frequency depends on regional regulations

³ Data rate and power consumption depend on Spreading Factor (SF); higher SF = longer range, lower rate, higher energy cost

These characteristics drive protocol selection based on deployment context and communication requirements:

Range vs. Throughput Trade-offs: LoRa provides 2-10 km range (with appropriate antenna configurations) but very limited throughput (0.3-50 Kbps), suitable for sparse periodic synchronization between distant partitions. Bluetooth Low Energy (BLE) and ESP-NOW offer moderate range (50-100 m) with higher throughput (1-2 Mbps), appropriate for dense local deployments where multi-hop forwarding is feasible and often required.

Discovery Mechanisms: Different protocols employ distinct peer discovery approaches. BLE advertising enables continuous passive discovery, ESP-NOW requires explicit peer lists, and ZigBee uses beacon-based association. μ -Babel's hybrid discovery mechanism (Section 3.6.3) will accommodate these differences while maintaining a unified peer view across protocols.

Energy Considerations: Protocol selection must account for energy costs: Wi-Fi consumes significantly more power than BLE, while LoRa power consumption depends heavily on SF configuration. The data compression mechanisms (Section 3.6.4) interact with protocol selection to minimize overall energy expenditure.

Together, these protocol characteristics illustrate the inherent trade-offs faced by resilient IoT deployments operating without reliable infrastructure. Understanding how range, throughput, discovery mechanisms and energy consumption vary across communication technologies is essential to establish an architecture that can adaptively leverage multiple protocols instead of relying on a single one.

2.1.3 Discussion

The communication technologies highlight that no single protocol can simultaneously optimize for range, throughput, energy efficiency, and autonomy under infrastructure failures. Instead, resilience emerges from the ability to combine the complimentary

features of each technology and adapt communication strategies to the specific context and requirements of a given environment.

These observations motivate an approach in which diversity is treated as a first-class requirement, and provide the technological context for the multi-protocol operation and architectural choices discussed in Chapter 3.

2.2 P2P Mesh Networking and Topology Management

Infrastructure-dependent star topologies in which devices communicate via a central coordinator or gateway suffer from a similar downside when compared with single-protocol communication: when that coordinator becomes unavailable or unreachable, the entire network loses connectivity.

This pattern pervades current IoT deployments, from Wi-Fi access point dependencies to LoRa Wide Area Network (LoRaWAN) gateway requirements, and becomes catastrophic in disaster scenarios where central coordinators are most likely to fail first, due to power outages, physical damage, or severe network congestion from increased emergency communications traffic.

FIXED: justify this claim

P2P mesh architectures address this limitation by distributing coordination across all participating nodes, thus eliminating single points of failure. Nonetheless, achieving robust mesh operation requires solving three interconnected challenges: neighbor discovery and establishment of initial connectivity, topology maintenance as nodes join/leave a network, and efficient data routing through multi-hop paths when direct communication becomes impossible or inefficient.

2.2.1 Peer Discovery and Topology Maintenance

As discussed in Section 2.3.2, the Multi-Protocol IoT Gateway implementation [23] demonstrates BLE-based neighbor discovery with Received signal strength indicator (RSSI) measurements for proximity estimation.

While that work focuses on multi-protocol integration, its topology management reveals a limitation of its tree-based approach: the system implements automatic parent reselection when coordinators fail, but the underlying tree structure imposes that nodes can only communicate through their parent-child relationships rather than arbitrary peer connections.

This restriction limits route diversity and resilience, creating dependency chains where a single intermediate node failure can disconnect entire subtrees.

A more sophisticated approach to membership management is presented in HyParView [25], in which each node maintains two distinct partial views for scalability: a small active view (size = fanout + 1) containing nodes with which symmetric links are

actively maintained, and a larger passive view serving as a backup pool of potential neighbors that may be promoted to the active view if one of its nodes fails.

The active view is managed reactively, such that nodes are added during join operations and removed upon failure detection, while the passive view is maintained cyclically through periodic shuffle operations that randomly exchange node identifiers between peers.

This hybrid strategy enables remarkable resilience, with the system recovering from 80% node failures with minimal reliability loss (maintaining 95% reliability) and from 50% failures in just 1-2 membership cycles, compared to 60+ cycles required by purely cyclic protocols like Cyclon [47].

HyParView's deterministic flooding approach, by broadcasting messages along the entire active view graph rather than probabilistic neighbor selection, enables fast failure detection since every active link is tested at each broadcast. The symmetric link requirement ensures bidirectionality: if node A can reach node B, then B can reach A, preventing the formation of one-way communication paths that complicate routing.

However, HyParView assumes TCP availability for maintaining persistent connections and using connection failures as implicit failure detectors. Additionally, it assumes that any node can communicate with any other node in its active view, which in practice requires some form of routing infrastructure.

This dependency on full network stack functionality makes direct application to resource-constrained embedded platforms challenging, though the architectural principles of hybrid views and shuffle-based passive view maintenance remain valuable.

The heterogeneous disaster IoT architecture [34] discussed in Section 2.3.1 employs Routing Protocol for Low-Power and Lossy Networks (RPL) for its Wireless Sensor Network (WSN) layer, demonstrating practical routing in resource-constrained disaster scenarios.

Their performance analysis reveals considerable trade-offs: convergence time scales linearly from 7 seconds for 20 nodes to 14.5 seconds for 100 nodes, while Packet loss ratio (PLR) at 10 hops reaches 80% with a 10-second sending interval but becomes acceptable at 20-second intervals.

These measurements highlight the throughput limitations imposed by tree topologies, in that their Instance 1 traffic (human data) must be restricted to 1-hop from the root node to ensure the least possible delay in its delivery, defeating the purpose of multi-hop mesh for critical communications.

The Delay-tolerant networking (DTN) component using mobile drones as data mules provides an alternative path for partitioned networks, with a maximum of 20 nodes per Destination-oriented directed acyclic graph (DODAG) to maintain acceptable convergence times during emergency situations, but this approach trades latency for eventual delivery rather than real-time mesh routing.

These approaches reveal fundamental trade-offs: tree-based protocols like RPL offer structured routing for resource-constrained devices but create single points of failure; fully distributed protocols like HyParView achieve resilience through P2P overlay management but assume network capabilities impractical for embedded platforms; and hybrid multi-layer approaches demonstrate practical deployment but still face throughput limitations.

This gap between resilience requirements and embedded capabilities motivates the exploration of topology management strategies more suitable for autonomous operation, particularly in unstable network conditions.

FIXED: add take-away for the section + set up narrative "motivates exploration..."

2.2.2 Routing in Partitioned and Intermittently Connected Networks

When network partitions prevent end-to-end paths, store-and-forward mechanisms enable eventual data delivery. The Bundle Protocol [38] addresses delay-tolerant networking through custody-based retransmission and opportunistic connectivity exploitation. While the protocol specification predates modern IoT deployments and was not designed specifically for resource-constrained devices, its core principles inform contemporary DTN approaches.

The framework presented in [30] implements elastic bandwidth utilization by dynamically adjusting transmission rates based on available connectivity, and supports scheduled, predicted, and opportunistic transmission windows, taking inspiration from the Bundle Protocol.

Data parcels are compressed, encrypted, and bundled before transmission, with a load balancer managing concurrent transfer threads to optimize bandwidth usage during brief connectivity windows.

While their Hypertext Transfer Protocol (HTTP)-based implementation targets cloud-backed IoT deployments, the core concepts of parceling data, maintaining transmission queues, and opportunistic forwarding during connectivity windows translate to P2P scenarios where aggregation nodes become neighbors in a mesh network.

2.2.3 Limitations of Centralized Coordination

The work on Resilient Edge-enabled IoT [9] addresses coordinator failures through dynamic leader election and backup mechanisms within their framework.

Their coordination model divides environments into collaboration areas, with resource-rich edge devices serving as coordinators that allocate tasks to workers under their supervision when problems arise. Workers, in turn, are *active agents* such as robots and IoT devices that reside in a particular environment, detect problems, and notify their coordinators.

When coordinators fail, the system automatically elects backups through adaptive decentralized consensus, providing "gentle degradation" during failures with restoration

after recovery. Multiple coordinators operate independently, eliminating single points of failure within the coordination model itself.

This architecture depends fundamentally on edge servers running JVM-based SCAFI middleware (a Scala library) to execute the aggregate programs that specify coordination behavior, and these heavyweight infrastructure requirements – both the Java runtime environment and resource-rich edge computing nodes – conflict with embedded platform constraints and infrastructure-failure scenarios.

While the aggregate computing paradigm separates concerns (sensing, actuation, communication, coordination), the implementation assumptions make it unsuitable for disaster-resilient systems where edge servers may be the infrastructure that fails. The formal guarantees of self-stabilization and compositional properties come at the cost of persistent computational infrastructure that an embedded-focused deployment cannot provide.

queremos minglé com o Babel... se calhar não entrar por aqui ou tudo bem porque não vamos usar os big-raspis para coordination?

2.2.4 Discussion

Existing peer discovery and topology maintenance approaches demonstrate mechanisms appropriate for distributed operation but not without their limitations for resource-constrained platforms aimed at disaster scenarios.

HyParView's [25] hybrid view architecture (small active, large passive) provides remarkable resilience to node failures, but the assumption of routing and TCP availability and persistent connections is not suitable for embedded devices using connectionless radio technology or operating under intermittent connectivity conditions.

The multi-protocol gateway's [23] BLE-based neighbor discovery works on our targeted hardware but imposes tree topologies with parent-child communication restrictions that limit route diversity and create dependency chains.

RPL routing in the heterogeneous disaster architecture [34] demonstrates practical WSN operation but has significant throughput limitations: 20-second minimum send intervals at 10 hops, and restriction of critical traffic to 1-hop from root nodes.

The store-and-forward mechanism presented in the elastic bandwidth framework [30] enables partition tolerance but targets cloud-backed deployments rather than P2P mesh scenarios.

Coordination frameworks like the one in [9] provide dynamic leader election but depend on JVM-based middleware on resource-rich edge servers.

We have found no work that integrates connectionless neighbor discovery, hybrid view maintenance, true P2P routing (not tree-based), and partition-tolerant store-and-forward on resource-constrained embedded platforms.

μ -Babel's approach (Section 3.6.2) must adapt HyParView's architectural principles to BLE/LoRa connectionless communication while supporting multi-hop routing without depending on coordinators or routing infrastructure.

2.3 Multi-Protocol Communication and Adaptive Protocol Selection

Single-protocol (i.e. Wi-Fi only, BLE only, LoRa only, etc.) communication architectures are susceptible to a fundamental vulnerability: if their chosen medium becomes unavailable or inefficient due to interference, range limitations, or infrastructure failures, the entire system loses connectivity and most times ceases to operate altogether.

This becomes especially grave in disaster scenarios where communication conditions change unpredictably due to factors such as Radio frequency (RF) interference caused by debris, obstacles blocking line-of-sight propagation, or damaged/unavailable access points.

Existing approaches to multi-protocol IoT systems can be divided into three categories: multi-channel resilience architectures that orchestrate communication technologies for emergency scenarios; implementations that show the feasibility of protocol heterogeneity on embedded platforms; and adaptive selection frameworks that switch protocols based on runtime conditions.

2.3.1 Multi-Channel Resilience Architectures

Several disaster-focused systems explicitly address infrastructure failures through the exploitation of protocol diversity.

The Always Connected Things (AWCT) framework [26] orchestrates Low-Power Wide Area Network (LPWAN) on top of LoRaWAN with ad-hoc networks (Bluetooth and Wi-Fi) specifically for standby emergency communication. Their architecture adds three modules to standard IoT devices (Raspberry Pi boards, in their test case): a battery module for power management, a power interrupt handler that triggers emergency mode when the power grid fails, and an ad-hoc bridge that forwards packets between the Bluetooth, Wi-Fi and LPWAN interfaces.

The system leverages dense IoT device deployment to provide emergency coverage, demonstrating that existing infrastructure can still serve a purpose during scenarios where the main power grid suffers issues.

However, AWCT's reliance on centralized LoRaWAN gateways for Internet connectivity creates a single point of failure when those gateways become unreachable.

A more comprehensive heterogeneous approach is presented in [34], which integrates High frequency (HF) radio with Near Vertical Incidence Skywave (NVIS), satellite links, WSNs, and DTN with mobile drones for disaster monitoring. Their system uses RPL [2] with three separate instances to differentiate traffic by priority: human data (voice/text via Bluetooth) receives the highest priority, followed by drone-collected data and finally sensor data.

The NVIS backhaul provides 250 km coverage radius without line-of-sight requirements, offering a cost-effective alternative to satellite communications.

While demonstrating successful real-world validation in Antarctica and urban deployments, the architecture's core NVIS topology with centralized coordination contrasts fully distributed operational requirements. Furthermore, their WSN layer requires a minimum 20-second sending interval to maintain acceptable packet loss rates at 10 hops, which highlights throughput limitations of single-channel tree topologies.

Security-focused multi-channel approaches like MCSC-WoT [6] combine Advanced Encryption Standard (AES) encryption with dynamic channel hopping across 2.4 GHz Wi-Fi channels to defend against jamming and eavesdropping attacks. Their lightweight synchronization mechanism minimizes energy consumption while maintaining security through Frequency-hopping spread spectrum (FHSS) patterns generated via Pseudorandom number generator (PRNG). Nodes that lose synchronization can rejoin by hopping to random channels and waiting for the next synchronization signal.

However, the system depends on a master node broadcasting synchronization signals and uses an initial PRNG seed shared among all nodes, raising questions about scalability and seed distribution mechanisms – particularly how new nodes can acquire seeds if deployed to a network some time after the system is operational.

2.3.2 Protocol Heterogeneity on Embedded Platforms

The work presented in [23] shows the technical feasibility of multi-protocol operation on commodity Wi-Fi/BLE modules by integrating ESP-NOW, ZigBee, and Modbus protocols on devices from the ESP32 family to construct multi-hop, tree-based wireless networks.

In this tree-based architecture, devices organize hierarchically with a central gateway serving as the root. Their implementation uses BLE advertising beacons for neighbor discovery with RSSI measurements (for distance estimation), computing parent selection priorities based on weighted combinations of child count, RSSI values, and hop count in the network tree of a given node. The system also supports automatic parent reselection when the current parent becomes unreachable, making it somewhat resilient to single node failures.

Testing demonstrated successful multi-hop operation up to 5 hops in office environments, but evaluation was limited to linear network topologies. Additionally, this architecture relies on a tree topology with parent/child communication that is centered around a gateway (coordinator) rather than being a P2P mesh, creating a single point of failure at the coordinator node.

2.3.3 Runtime Adaptive Protocol Selection

The MINOS platform [45] exemplifies the limitations of centralized multi-protocol approaches. While providing sophisticated multi-protocol support (CORAL-SDN and Adaptable-RPL) with dynamic protocol deployment and real-time parameter tuning, the system depends fundamentally on a centralized SDN controller, MQTT broker, and web server infrastructure. The architecture's single point of failure means that when the controller becomes unreachable the entire system loses its adaptive capabilities and reverts to static operation at best, or complete failure at worst.

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2.3.4 Discussion

Existing frameworks demonstrate the technical feasibility of heterogeneous channel/protocol coordination but rely on centralized control mechanisms incompatible with disaster scenarios. AWCT [26] depends on centralized LoRaWAN gateways for Internet connectivity, the heterogeneous disaster architecture [34] uses centralized NVIS coordination with minimum 20-second send intervals at 10 hops, MCSC-WoT [6] requires master nodes broadcasting synchronization signals, the multi-protocol gateway [23] uses tree topologies with gateway coordinators as single points of failure, and MINOS [45] assumes SDN controllers with persistent MQTT/web infrastructure.

We have found no work that addresses fully distributed multi-protocol coordination where nodes can autonomously select and switch protocols without resorting to persistent infrastructure. μ -Babel's approach to decentralized protocol selection is detailed in Section 3.6.3 along with masterless synchronization mechanisms.

2.4 Decentralized Synchronization and Time Coordination

Multi-channel communication strategies, particularly those that employ coordinated channel hopping for security or efficiency reasons, require nodes to maintain synchronized clocks. Without time synchronization, devices cannot agree on when to switch channels, rendering coordinated multi-protocol operation unfeasible.

Traditional master-based synchronization approaches – where a designated coordinator is responsible for communicating timing signals – once more fall into the limitation of having a single point of failure in that same master node: should it fail, the entire network becomes susceptible to losing accurate temporal reference, leading to the collapse of any coordination efforts.

This is a notorious challenge to overcome when designing distributed systems, and it becomes more prominent in the disaster scenarios discussed thus far, where infrastructure failures are likely to eliminate the nodes responsible for time synchronization in a given system.

Achieving robust time synchronization in such situations requires fully decentralized approaches that eliminate central coordinator dependencies while still remaining sufficiently lightweight to be executed on resource-constrained platforms.

2.4.1 Master-Based Synchronization as Counterexample

As discussed in Section 2.3.1, the MCSC-WoT framework [6] demonstrates a security-focused multi-channel hopping approach aimed at embedded platforms, but with a fundamental dependency on a master node broadcasting synchronization signals.

This approach uses a shared PRNG seed to generate channel-hopping sequences, with the coordinator node broadcasting periodic synchronization beacons that participant nodes use to compensate for clock drift. Nodes that lose synchronization altogether can rejoin by hopping to a random channel and waiting for the next coordinator beacon.

While the proposed goals were achieved in terms of minimizing energy expenditure and maintaining FHSS patterns, it inherits the fundamental limitation of all master-based approaches mentioned at the start of Section 2.4. If the master fails or becomes unreachable, participating nodes gradually drift out of synchronization until coordinated channel hopping is no longer possible.

The system's clock drift compensation algorithm demonstrates the feasibility of synchronization on ESP32 platforms, and their measured AES encryption performance validates that lightweight security can coexist with time synchronization on resource-constrained devices. However, the architectural dependency on a central coordinator fundamentally conflicts with infrastructure-independent operation requirements we aim to fulfill.

2.4.2 Gossip-Based Masterless Synchronization

Decentralized time synchronization eliminates coordinator dependencies by having nodes reach consensus on clock values through distributed local interactions. Two complementary approaches demonstrate the viability of gossip-based synchronization for WSNs.

The Randomized gossip-consensus-based sync (RGCS) algorithm proposed in [50] addresses time synchronization in dynamic WSNs through randomized asynchronous gossip. Each node maintains a logical clock (T) composed of rate (α) and offset (β) parameters, which together transform the node's hardware clock (τ) into synchronized logical time.

Rather than requiring fixed communication links between specific node pairs which might be fragile in dynamic topologies, their approach uses Poisson-based randomized link activation where each potential synchronization link activates with intensity λ .

The synchronization process operates through pairwise gossip exchanges: when a link activates, the triggering node sends a Sync-L beacon selecting a triggered neighbor, followed by bidirectional exchange of multivariable messages containing each node's current logical clock parameters $[\alpha, \beta, \tau]$.

The asynchronous randomized timing of these exchanges is advantageous in that collision rates drop to near zero compared to 19-23% for deterministic communication protocols, as independent Poisson intervals make simultaneous transmissions to the same receiver statistically unlikely.

The proposed RGCS employs a converge-to-max criterion rather than average-value consensus. This maximum-based approach achieves finite time convergence significantly faster than average-based protocols that require many iterations to converge under significant clock drift. Offset compensation follows suit, with nodes adjusting β parameters based on the difference between their and the neighbors' logical clocks.

Bounded communication delays are handled through a least-square estimation low-pass filter. This addresses the realistic concern of uplink and downlink delays differing, and avoids the symmetric delay assumptions made by many theoretical protocols. The filter's weighing parameter decreases over time, in order to restrain the negative effects of additive noise in stochastic approximation.

Storage complexity remains $O(|N_i|)$ per node per iteration (proportional only to the number of neighbors, not network size) thus ensuring scalability. The protocol simultaneously compensates both clock rate and offset, unlike approaches that handle these separately and thus require additional convergence time.

2.4.3 Coordination Through Passive View Maintenance

As discussed in Section 2.2.1, the shuffle-based passive view maintenance presented in HyParView [25] provides a complementary coordination mechanism. While primarily designed for topology management, the periodic shuffle operations in which nodes exchange lists of known peers can also provide a *catalog* of potential synchronization partners beyond their immediate active neighbors.

Nodes performing shuffle exchanges already communicate periodically; these same communication windows can opportunistically carry synchronization messages (piggy-backing), reducing protocol overhead. The passive view serves as a pool of potential sync partners, so that when a node's active sync neighbors become unreachable, it can initiate sync exchanges with passive view members, providing added resilience to topology changes without requiring global network knowledge.

2.4.4 Discussion

Master-based synchronization approaches like MCSC-WoT [6] demonstrate feasibility on ESP32 platforms with measured AES encryption coexistence, but create single points of failure when master nodes become unreachable.

RGCS [50] eliminates coordinator dependencies, and provides converge-to-max synchronization with Poisson-based randomized gossip that achieves near-zero collision rates with just $O(|N_i|)$ storage complexity per node. However, it assumes homogeneous

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sion

single-protocol networks where all nodes use the same communication medium with identical energy characteristics and range properties.

HyParView's [25] shuffle-based passive view maintenance provides complementary coordination through periodic peer exchanges that could carry piggybacked sync messages, but was not designed for time synchronization integration. One example would be when a node receives a shuffle message containing peer descriptors (IDs and capabilities): it can evaluate these peers as candidates for time synchronization based on their advertised characteristics – for instance, prioritizing peers with access to Network Time Protocol (NTP), Global Positioning System (GPS) clock synchronization, or simply those maintaining more reliable clock sources.

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We have found no work that addresses decentralized time synchronization across heterogeneous multi-protocol networks where different communication technologies (BLE, LoRa, Wi-Fi) have distinct energy costs, range, and reliability properties that warrant protocol-specific gossip rates.

μ -Babel's approach (Section 3.6.3) will strive to extend RGCS to multi-protocol scenarios with per-protocol Poisson processes while preventing protocol selection oscillations during sync events.

2.5 Lightweight Data Compression and Energy Efficiency

Energy constraints present one of the most fundamental limitations in battery-powered IoT deployments, particularly in hazardous deployments where replacements and recharging are impractical, and during disaster scenarios where grid power likely becomes unavailable.

While multi-protocol communication and decentralized coordination efforts provide resilience, they also introduce increased energy costs through frequent transmissions, protocol adaptation overhead and forwarding costs.

Data transmission dominates energy consumption across wireless technologies: LoRa and SIGFOX can represent up to 99.9% of power consumption for transmission, while even shorter-range technologies like BLE and IEEE 802.15.4 dedicate 85-90% of their energy budget to radio operations [43].

This reality makes data reduction techniques essential for extending operational lifetime in infrastructure-independent scenarios. However, compression itself consumes energy through additional computational work, thus creating an unavoidable tug-of-war between compression overhead and transmission benefits.

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2.5.1 Prediction-Based Data Reduction

The Ambrosia protocol [43] demonstrates that lightweight prediction can achieve substantial data reductions on resource-constrained devices. The core idea is that not all sensor

readings need to be transmitted if their values can be accurately predicted at the server (within the bounds of a specified error threshold), provided that both sender and receiver maintain a synchronized "prediction state".

Their approach uses window-based forecasting where the next sample is predicted by adding the average differences between recent previous samples, making it dramatically lighter than sophisticated time-series forecasting techniques like Autoregressive integrated moving average (ARIMA) [40] while still achieving comparable data reduction performance.

The protocol works by sending the first w (window size) true samples collected by the sensor node to the server, and for every sample after that comparing its true value to the one predicted locally; the true sample value is sent to the server only if the absolute difference between it and the prediction is greater than a user or application-specified error threshold δ .

The same simple prediction scheme is used on both ends of this communication, and the predicted value is the one used for further predictions – even on the sensor node that knows the true value, to ensure consistency between the two endpoints.

This design achieves up to 60% data reduction with appropriate δ configuration while maintaining sufficient accuracy for diverse applications: for error-tolerant use cases like anomaly detection, δ values will be higher to allow for reasonable fluctuations, thus enabling more significant data reductions, while error-sensitive applications requiring higher precision will naturally lean towards stricter error thresholds, but still benefitting from some amount of data reduction.

The major takeaway from this work is that the window-based prediction executed 99% faster than ARIMA forecasting in their evaluation, making it viable for streaming sensor data at high rates. The approach achieved 2× battery lifetime improvement in high-traffic scenarios and demonstrated compatibility with extremely constrained devices (livestock ear tags with Atmel 8-bit microcontrollers).

2.5.2 Lightweight Compression

While prediction-based reduction minimizes the number of transmissions, compression techniques are essential to reduce the size of the data that *must* be transmitted.

A few complementary approaches address different considerations and constraints: hybrid schemes balance accuracy with transmission costs through lossy/lossless models; lossless time-series compression exploits the natural temporal relation in continuous sensor readings; and two-tier architectures apply different techniques at both the sensor and gateway level to reduce energy costs.

Hybrid Lossy/Lossless Compression

The hybrid compression scheme presented in [13] addresses accuracy requirements by separating real-time lossy transmission from on-demand lossless reconstruction. The Fan

algorithm adaptively sub-samples data maintaining bounded error ε , achieving average $7.8\times$ and $2.1\times$ Compression ratio (CR) for lossy and lossless compression, respectively.

The approach provides graceful degradation through battery-aware switching, in that when battery drops below a certain threshold (e.g., 20%), the system switches to lossy-only transmission instead of lossless, extending lifetime at the cost of reconstruction accuracy.

This trade-off aligns with disaster-scenario requirements where sustained operation during prolonged periods of network outages often takes higher priority over perfect data-fidelity, as approximate sensor readings over extended periods of time provide greater situational awareness than high-precision measurements from devices that quickly deplete their batteries. However, safety-critical sensors (e.g., structural integrity sensors, hazardous gas detectors) may require different thresholds (if any) or other policies to balance longevity with accuracy requirements.

Lossless Time-Series Compression

Sprintz [7] targets lossless compression for multivariate integer time-series with extreme resource constraints, achieving 2-10 \times compression ratios with <1KB memory footprint and 8-sample block sizes. The four-component algorithm combines forecasting (delta coding or Fast Integer REgression (FIRE) online learning), bit-packing with zigzag encoding, Run-length encoding (RLE) for all-zero blocks, and optional Huffman coding.

Delta coding is extremely fast, and when combined with RLE becomes even more so, as it yields a run of zero errors if the data is constant, which is likely to happen for nominal sensor readings. FIRE forecasting yields better compression, but its online learning approach place it beyond the scope of this paper.

A *forecaster* is employed to predict each sample based on previous ones, and the difference between the next and the predicted sample are encoded. This difference is typically closer to zero than the next sample itself.

Any prediction errors from the previous forecasting step are zigzag encoded as a "payload", and a header is prepended with sufficient information to invert the bit packing.

If a block happens to consist only of zeros, this approach waits for a block in which there is a non-zero error, and the number of all-zero blocks are written out using RLE instead of the (empty) payload.

Finally, the bit packed representation of each block can be *entropy coded* using a Huffman coder, applied to the headers and payloads. This is done after bit packing because not only is it faster, but it also increases compression.

Decompression can achieve up to 3GB/s throughput without Huffman coding (>500MB/s with Huffman) in a single thread, enabling high-speed data retrieval on gateway-tier devices. Compression maintains >200MB/s on 8-bit data, sufficient for real-time operation even on embedded platforms.

Two-Tier Compression Architecture

The two-tier data reduction technique proposed in [35] applies compression at both sensor nodes (Tier 1) and gateways (Tier 2) to reduce energy consumption in the overall system.

Sensor nodes employ Delta encoding followed by RLE, allowing this approach to achieve 80-84% compression by exploiting the temporal correlation in sensor data. Gateways perform hierarchical clustering based on the Minimum description length (MDL) principle – that is, combining pairs of datasets into one cluster when their combined representation is shorter than separate encodings – to transmit *hypothesis* data sets and difference vectors instead of full data sets.

The Delta+RLE implementation – validated through OMNeT++ simulation – proves to be lightweight enough for resource-constrained nodes, while gateway clustering reduces transmission to cloud infrastructure.

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2.5.3 Receiver-Side Energy Considerations

An often overlooked aspect of IoT energy optimization is the asymmetry between sender and receiver energy consumption. The adaptive protocol selection framework analysis [53] revealed that receiver nodes consistently consume 15-20% more energy than senders across all evaluated protocols (MQTT, CoAP, HTTP).

This asymmetry shifts the energy bottleneck from endpoint sensors to intermediary nodes like gateways and has direct implications for gateway power management in disaster scenarios where battery replacement becomes impossible. The increased receiver cost stems from the added computational work required for parsing, processing, and managing incoming messages, particularly at high message rates.

For μ -Babel's proposed heterogeneous architecture where sensor nodes might eventually forward data to aggregators/gateways, this finding has direct implications in that managing the power consumption of these gateways becomes as significant as sensor power optimization, since one gateway might receive substantial data from several nodes even if individually they don't perform frequent transmissions.

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2.5.4 Discussion

Existing data reduction techniques demonstrate substantial energy savings on resource-constrained platforms but assume single-protocol deployments.

Ambrosia's [43] window-based prediction achieves 60% data reduction with 99% faster execution than ARIMA while maintaining accuracy within application-specific error thresholds, validated on 8-bit microcontrollers.

Sprintz [7] provides lossless time-series compression with 2-10× compression ratios using <1KB memory and 8-sample blocks, combining delta coding, zigzag encoding, RLE, and optional Huffman coding with >200MB/s compression throughput suitable for real-time embedded operation.

Hybrid lossy/lossless [13] schemes enable graceful battery-aware degradation, while two-tier architectures [35] (Delta+RLE at sensors, MDL clustering at gateways) achieve 80-84% compression exploiting temporal correlation.

None of these approaches address protocol heterogeneity by themselves, and different wireless technologies have distinct energy costs and range/accuracy trade-offs that warrant protocol-specific compression/reduction strategies.

We have found no work that integrates prediction-based reduction with protocol-aware adaptive thresholding and compression tuning for heterogeneous multi-protocol networks.

μ -Babel's approach (Section 3.6.4) should, therefore, consider lightweight prediction and compression in multi-protocol scenarios with per-protocol thresholds and strategy selection while accounting for gateway receiver energy costs.

2.6 Security Considerations and Scope Delimitation

This thesis focuses on resilience and decentralization at the physical and link layers, and, as such, security mechanisms operating at higher layers (i.e., authentication, key management, privacy-preserving mechanisms) are considered out of scope for the present body of work.

Security and privacy are nonetheless critical concerns in emergency and disaster-focused scenarios, where sensitive data such as location tracking and environmental conditions could be the target of malicious actors seeking to disrupt system operation, inject false information, or compromise user safety.

Substantial research effort has been devoted to addressing these concerns in resource-constrained IoT systems. While a comprehensive discussion of such mechanisms is beyond the scope of this work, a brief overview of relevant approaches is provided below to contextualize potential future extensions of the proposed μ -Babel framework.

Several lightweight authentication and key management approaches demonstrate that cryptographic operations based on hash functions, symmetric encryption and polynomial secret sharing are feasible on embedded platforms. The Group Authentication Scheme at the Edge (GASE) [32] relies on aggregated Message Authentication Codes (MACs) and threshold techniques to reduce computational overhead, while the Lightweight Polynomial-based Key Management (LPKM) protocol [17] enables scalable and pairwise key establishment with low storage requirements and distributed revocation. Symmetric encryption – notably AES – has been proven to be practical on ESP32 devices without prohibitive performance or energy costs, as demonstrated by the MCSC-WoT framework [6].

Other lines of work approach privacy-preservation by relying on hardware fingerprints and pseudonym systems, including protocols based on Physically Unclonable Functions

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(PUFs) [20, 51, 11]. However, these approaches assume specialized hardware or centralized verification infrastructure, greatly limiting their applicability to commodity devices and infrastructure-less deployments. Similarly, zero-preloading key agreement schemes [27] attempt to eliminate prior trust establishment but face unresolved challenges related to authentication bootstrapping and revocation.

Several proposals focus instead on blockchain technology or distributed ledgers to address IoT security concerns, including access control [46] and decentralized Public key infrastructures (PKIs) [48]. While these approaches remove single points of failure, they assume persistent connectivity incompatible with disaster scenarios and computational capabilities far beyond those available in low-end embedded platforms; pairing-based decentralized encryption schemes further exacerbate this issue [22].

Ultimately, existing work confirms the feasibility of lightweight security mechanisms on constrained devices, but also highlights a persistent reliance on infrastructure availability or pre-established trust. Rather than addressing these challenges directly, the proposed μ -Babel framework focuses on providing a foundation for resilient and decentralized communication on top of which security mechanisms may be integrated.

2.7 Summary

Table 2.2 summarizes the key architectural differences between existing work and μ -Babel's planned vision across the technical domains discussed in this chapter.

Table 2.2: Related work summary

Domain	Related Work Limitations	μ -Babel Approach
Protocol coordination	Master nodes (MCSC-WoT), SDN controllers (MINOS), gateway coordinators (multi-protocol gateway)	Fully distributed selection
Mesh topology	Tree-based restrictions, TCP-dependent links (HyParView), centralized coordination	Connectionless BLE/LoRa discovery, hybrid views, true P2P routing
Time sync	Master beacons (MCSC-WoT), single-protocol gossip (RGCS)	Protocol-specific Poisson rates, multi-graph converge-to-max
Compression	Single-protocol optimization (Ambrosia, Sprintz), sender-focused	Protocol-aware adaptive thresholds, gateway receiver energy management

A common pattern is present across these domains: existing approaches rely on centralized coordination mechanisms that become single points of failure during infrastructure collapse.

Protocol coordination depends on master nodes or SDN controllers, mesh topologies assume consistent connectivity or gateway coordinators, time synchronization requires

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master beacons, and authentication protocols depend on centralized servers or group leaders.

Even distributed approaches like HyParView and RGCS make assumptions (routing infrastructure, persistent TCP connections, single-protocol homogeneity) incompatible with resource-constrained multi-protocol disaster scenarios.

In the next Chapter we discuss how these limitations inform the design of μ -**Babel** as motivating factors for developing a fully decentralized architecture – one that prioritizes autonomous operation, protocol diversity, and resilience under unstable connectivity.

There we highlight the system requirements, operating conditions, and architectural choices that will enable robust communication and coordination across heterogeneous devices.

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SOLUTION ARCHITECTURE

3.1 Recap

3.1.1 The Problem

Current Internet of Things (IoT) and domotics systems assume continuous infrastructure availability, an assumption that proves rather catastrophic during disasters, in hazardous environments, or when privacy requirements demand local autonomy (Chapter 1).

The widely-adopted, cloud-centric architectural model creates single points of failure in myriad ways: devices depend on remote servers for control logic, local access points for Internet connectivity, and designated coordinators for synchronization.

When infrastructure fails, whether due to large-scale outages [44] or disaster-related network collapse, coordinated operation becomes impossible despite the local availability of physical computational resources.

3.1.2 Technical Challenges

Throughout Chapter 2 we identified the core challenges that prevent existing approaches from fully addressing the overarching problem of reliable autonomous operation:

Centralized coordination: Protocol selection requires master nodes or SDN controllers. Topology management depends on gateway coordinators. Time synchronization relies on master beacons. Authentication requires persistent servers or group leaders. When these central elements fail or become unreachable, coordination collapses despite local devices remaining operational.

Single-protocol operation: Gossip-based synchronization assumes homogeneous networks where all nodes share identical communication characteristics. Compression techniques optimize for single technologies without considering protocol-specific energy costs. Topology maintenance protocols assume either persistent connections or uniform radio properties ill-suited for multi-protocol embedded platforms.

Infrastructure dependencies: Blockchain-based security requires connectivity to distributed ledgers. Edge computing assumes eventual cloud access. Hybrid architectures distribute processing but remain dependent on infrastructure for coordination. Systems designed with "infrastructure as fallback" still fail when that fallback becomes unavailable indefinitely.

Resource constraints: Sophisticated coordination mechanisms exceed embedded platform capabilities. Security schemes require specialized hardware (Physically Unclonable Functions (PUFs)) or computationally expensive operations (bilinear pairings). The tension between autonomous operation demands and limited computational, memory, and energy budgets remains unresolved in existing approaches.

μ -**Babel** will address these interconnected challenges through a unified framework designed specifically for resource-constrained multi-protocol operation during infrastructure failures. The following sections detail the design goals, targeted platforms, and design decisions that enable autonomous coordination without central dependencies.

3.2 System Requirements

3.2.1 Overview of μ -Babel's Approach

In order to effectively overcome the aforementioned challenges, we can highlight a set of functional requirements that μ -**Babel** must meet in order to eliminate the *dependency* on centralized coordination, continuous connectivity, and cloud infrastructure; and a set of non-functional requirements which would strengthen the overall quality of the framework and ensure it can be widely adopted with ease.

Functional Requirements

Infrastructure Independence: The system must maintain communication capabilities when traditional infrastructure becomes unavailable. Rather than treating this unavailability as an exceptional condition requiring failover mechanisms, our architecture will focus on autonomous Peer-to-peer (P2P) connectivity as the baseline mode of operation, with infrastructure integration as an added benefit when available, rather than a hard dependency.

Autonomous Operation: Devices will self-organize into functional networks, dispensing centralized coordinators and static master nodes. Instead, local P2P gossip interactions will be the basis for coordination and synchronization.

"something something these are essential ..." ou o parágrafo anterior basta?

Multi-Protocol Exploitation: Communication strategies will adapt to the available mediums and device resources without centralized control. Communications will continue despite disruption of individual protocols through adaptive selection and opportunistic protocol switching. Protocol diversity provides resilience through redundancy, and this should be achieved with minimal impact on the upper layers of the application (through abstractions).

Scalability: The system must support deployments ranging from small, localized networks to large-scale, highly distributed collections. Rather than requiring global knowledge of all participants, nodes organize into local groups where elected leaders maintain state only for their immediate members. This area-based approach – analogous to Open Shortest Path First (OSPF)’s autonomous systems [31] – distributes coordination load and enables networks to scale while keeping per-node overhead bounded by local group size.

Resource Efficiency: Embedded platforms inevitably present strict resource constraints (i.e., memory storage, processing, battery), thus all protocols and algorithms must execute within these limitations. In order to enable this, lightweight data compression and reduction techniques will be employed to make the most out of the limited storage and battery available, and adaptive protocol selection will take into account device capabilities and available resources.

Non-Functional Requirements

Optional Cloud Integration: When infrastructure *is* available, it should be taken advantage of. Thus, μ -Babel will exploit it for tasks that are beyond the capabilities of individual nodes or even gateways, such as data aggregation, analytics, and long-term storage. Cloud integration thus becomes an optional – and welcome – addition to the repertoire of features, but does not itself become a requirement for dependable operation.

Interoperability: Following optional cloud integration, widely used IoT protocols such as MQTT and Constrained Application Protocol (CoAP) can be leveraged to enable integration with existing IoT platforms and services when conditions allow.

3.3 Operating Conditions and Failure Modes

μ -Babel is designed to remain operational under conditions where traditional infrastructure becomes unreliable or unavailable. This section establishes the operational environment and failure scenarios that drive our architectural decisions.

"something something these are not essential ..." ou o parágrafo inicial basta?

3.3.1 Target Operating Environments

μ -Babel targets deployments where infrastructure availability cannot be guaranteed, and in which autonomous operation is a necessity:

Disaster Scenarios: Locations susceptible to natural disasters (i.e., earthquakes, floods, fires) or infrastructure failures that eliminate power grids and network access points.

Remote or Hazardous Locations: Industrial sites, wilderness monitoring, or dangerous environments where manual intervention and infrastructure maintenance are impractical or impossible.

Local-First Deployments: Environments where data sovereignty requirements or privacy concerns require local processing and control rather than cloud dependency, even when connectivity is available.

3.3.2 Expected Failure Modes

The architecture must maintain functionality under the following conditions:

Infrastructure Collapse: Complete loss of access points and external network connectivity. Devices must self-organize and maintain P2P communication without infrastructure support.

Communication Medium Disruption: Individual protocols may become unavailable due to interference, jamming, congestion, or physical obstacles blocking signal propagation. Multi-protocol diversity provides alternative communication paths and ensures continued operation.

Network Partitioning: Physical damage or communication range limitations may split networks into isolated subgroups. Devices within partitions must continue autonomous operation while supporting eventual reconciliation when connectivity is restored.

Node Failures: Individual devices may become unavailable due to battery depletion, physical damage, or environmental conditions. The network must maintain operation despite individual node failures without depending on specific coordinator nodes.

Intermittent Connectivity: Communication links may be unstable with varying packet loss, latency, and availability. Protocols must tolerate unreliable links and adapt to changing conditions.

move this paragraph to use cases later?

3.4 Hardware Platform

μ -**Babel** targets resource-constrained embedded platforms commonly used in IoT and domotics systems. The following sections cover device specifications and communication protocol characteristics, establishing the foundation for our multi-protocol, heterogeneous architecture.

We distinguish between two device classes based on their capabilities and roles within the architecture.

3.4.1 Class-1: Battery-Powered Sensor Nodes

The primary targets for μ -**Babel**. These embedded devices handle sensing, actuation, P2P communication, and decentralized coordination without requiring external infrastructure.

Table 3.1 presents the resources available across Class-1 devices. These span from basic sensor nodes (Pico 2W, ESP32-C5/C6) to more capable embedded devices (ESP32-S3/P4), enabling role differentiation within the autonomous peer network.

TODO: figure out table spacing

Table 3.1: Class-1 Device Specifications

Device	CPU ¹	CPU Clock	SRAM	PSRAM ²	Flash
Raspberry Pi Pico 2 W	Dual-Core Arm Cortex-M33	150 MHz	520 KB	✗	4 MB
ESP32	Dual-Core Xtensa LX6	240 MHz	520 KB	✗	16 MB
ESP32-C5	Single-Core RISC-V	240 MHz	384 KB	✗	4 MB
ESP32-C6	Single-Core RISC-V	160 MHz	512 KB	✗	4 MB
ESP32-S3	Dual-Core Xtensa LX7	240 MHz	512 KB	8 MB	16 MB
ESP32-P4	Dual-Core RISC-V	360 MHz	768 KB	32 MB	32 MB

¹ All supported devices are 32-bit

² Pseudostatic random-access memory (PSRAM) is supported on all devices, but not available in the particular models used during development

Resource availability within Class-1 devices varies significantly both in processing power and available memory. While there are no fixed roles, capability awareness naturally influences dynamic selection procedures:

Data Aggregation and Buffering: Higher-resource devices (ESP32-S3/P4) can maintain larger message buffers and aggregate data from multiple sensor nodes, which is particularly valuable during network partitions when store-and-forward mechanisms come into effect.

Protocol Bridging: Devices with broader protocol support naturally serve as bridges between protocol domains, forwarding messages across different mediums based on advertised capabilities.

Protocol Diversity and Device Coverage

Table 3.2 shows the communication protocols supported across Class-1 devices, demonstrating considerable heterogeneity.

Table 3.2: Supported Communication Protocols

Device	Wi-Fi	Bluetooth	ESP-NOW	ZigBee	LoRa ¹
Pico 2W	2.4GHz	Classic + Low Energy (LE) 5.2	✗	✗	✓
ESP32	2.4 GHz	Classic + LE 4.2	✓	✗	✓
ESP32-C5 ²	2.4/5 GHz	LE 6.0	✓	3.0	✓
ESP32-C6 ²	2.4 GHz	LE 5.3	✓	3.0	✓
ESP32-S3	2.4 GHz	LE 5.0	✓	✗	✓
ESP32-P4 ³	2.4 GHz	LE 5.3	✓	✗	✓

¹ Via external SX1276/SX1262 transceiver, controlled through Serial Peripheral Interface (SPI)

² ESP32-C5/C6 also support 802.11ax (Wi-Fi 6) and Thread, but we focus on ZigBee given its wider adoption in IoT systems

³ ESP32-P4 does not possess wireless capabilities by itself and relies on a C6-Mini coprocessor, with ZigBee support still in development

Protocol support across devices provides several advantages:

Coverage: All devices support at least three protocols (Wi-Fi, Bluetooth Low Energy (BLE), LoRa), ensuring baseline multi-protocol operation across the entire platform range. Even the least capable device (Pico 2 W) can participate in protocol-diverse mesh networks.

Protocol-Specific Advantages: ESP32 family devices can leverage ESP-NOW for P2P communication without infrastructure. ESP32-C5/C6 additionally provide ZigBee support for standardized mesh networking – particularly interesting in building automation and domotics contexts.

Notably, ESP32-C5 supports ESP-NOW over both 2.4 and 5 GHz bands, which provides a crucial short-range alternative path for congested environments where BLE, Wi-Fi and ZigBee all contend for 2.4 GHz channels.

Long-Range Capabilities: All platforms support external LoRa transceivers via SPI, enabling long-range communication for disaster scenarios. This addresses the multi-protocol resilience requirements discussed in Section 2.3.

These factors combine to provide multi-protocol resilience: spectrum diversity across 2.4/5 GHz and sub-GHz bands, infrastructure-free operation via ESP-NOW and BLE, and long-range connectivity through LoRa.

Together, these Class-1 specifications and multi-protocol capabilities establish the core hardware foundation for μ -Babel’s autonomous operation. Section 3.6 details how these resources are exploited in the proposed architecture.

3.4.2 Class-2: Optional Aggregation Nodes

Resource-rich devices that run the broader Babel [18] framework and are mentioned here for the sake of integration with broader systems, but are not required for μ -Babel's core autonomous operation.

This class encompasses general-purpose computing devices such as desktops, laptops, and servers. For development purposes, and staying within the realm of embedded platforms, we specifically utilize Raspberry Pi Single-board computers (SBCs), whose specifications are available in Table 3.3.

Table 3.3: Class-2 Device Specifications

Device	CPU ¹	CPU Clock	RAM ¹	Storage
Raspberry Pi 5	Quad-Core 64-bit Arm Cortex-A76	2.4 GHz	4-16 GB	Variable external
Raspberry Pi 4	Quad-Core 64-bit Arm Cortex-A72	1.8 GHz	4-8 GB	Variable external
¹ Configurations with less Random-access memory (RAM) are available but were not considered				

Class-2 devices serve as *optional* infrastructure that, when present, can perform: (1) Data aggregation and long-term storage; (2) Computationally intensive analytics beyond Class-1 capabilities; and (3) Optional cloud bridging when external connectivity is available and desired (e.g., MQTT brokerage, CoAP endpoints).

Critically, these devices *do not* serve as coordinators or single points of failure. μ -Babel networks will operate fully autonomously with Class-1 devices alone; Class-2 nodes simply participate as peers that happen to have greater resources. Their absence does not prevent network formation, peer discovery, or autonomous operation.

3.5 Software Stack and Development Environment

μ -Babel is implemented on top of well-established embedded software platforms that provide hardware abstraction and concurrency support while remaining sufficiently lightweight for resource-constrained platforms.

The system leverages vendor-supported Software Development Kits (SDKs) and a minimal Real-time operating system (RTOS) to ensure portability across Class-1 devices while maintaining fine-grain control over hardware resources.

3.5.1 Operating System Layer

FreeRTOS [4] serves as the RTOS across all Class-1 devices. It provides deterministic task scheduling and inter-task communication implemented around a single queue primitive [5]. This queue mechanism forms the basis for queues (as data structures), semaphores, and mutexes. FreeRTOS also allows for configurable memory allocation and per-task stack management, all without the overhead of a full-featured Operating system (OS).

This allows multiple protocol components and I/O handlers to execute concurrently while preserving predictable timing behavior, which is essential for networking and radio-driver applications.

FreeRTOS is the *de facto* OS on ESP32-family devices, and is easily brought into the Pico platform. By using it as a common execution runtime, μ -**Babel** maintains a consistent concurrency and timing model across heterogeneous microcontroller platforms, simplifying both portability and implementation.

3.5.2 Platform SDKs and Hardware Abstractions

Hardware-specific functionality is accessed through vendor abstractions. The Espressif IoT Development Framework (ESP-IDF) [15] is used for the ESP32 family, while the Raspberry Pi Pico SDK [36] is used for RP2350-based devices.

These frameworks provide Application programming interfaces (APIs) for peripheral access (e.g., General-purpose input/output (GPIO), SPI, Inter-Integrated Circuit (I²C), Universal asynchronous receiver-transmitter (UART)), timers, interrupt handling, Wi-Fi/BLE radio stacks, and – where available – hardware-accelerated cryptographic operations.

By encapsulating platform-specific bare-metal details behind stable interfaces, they allow μ -**Babel** to target multiple microcontrollers through a set of thin Hardware Abstraction Layers (HALs), enabling re-utilization of higher-level protocols and coordination logic across Class-1 devices. μ -**Babel** is implemented in C, as it allows precise control over memory usage, interrupt behavior, and timing, which are essential in embedded environments with limited memory strict real-time constraints.

3.5.3 Concurrency Model

μ -**Babel** follows a task-based concurrency model built on FreeRTOS primitives. Protocol handlers execute as independent tasks, allowing computation, I/O, and radio operations to execute concurrently and predictably. Inter-task communication is handled through message queues, while mutexes and semaphores are used to protect shared state and coordinate access to hardware resources.

An event dispatcher provides higher-level coordination by routing events between protocol components and system services. This structure decouples individual components, simplifies reasoning about system behavior, and supports incremental extension of the protocol stack.

3.6 System Model

μ -Babel is structured as a layered architecture where components interact to achieve autonomous multi-protocol operation. This section presents the key architectural components organized into four functional layers: Core Infrastructure, Discovery & Topology, Coordination, and Data Management.

Figure ?? illustrates the overall component organization and interactions.

3.6.1 Core Infrastructure Layer

The Core Infrastructure Layer provides the foundational services for protocol management, event dispatching, and connection handling. These components abstract protocol heterogeneity and provide unified interfaces to higher layers.

Protocol Manager

The Protocol Manager maintains runtime information for all available communication protocols and device capabilities. It registers available protocols during system initialization and tracks their operational status throughout device operation.

Key responsibilities include:

- Registering protocol handlers and their characteristics (range, throughput, energy cost)
- Tracking which protocols are supported on each device based on hardware capabilities
- Enabling or disabling protocols dynamically based on runtime conditions (e.g., battery level, interference detection)
- Providing protocol capability information to the Protocol Selector for adaptive selection decisions

Event Dispatcher

The Event Dispatcher implements an event-driven coordination model that decouples components and enables asynchronous communication between services. Components register interest in specific event types and subtypes, and the dispatcher routes incoming events to all registered subscribers.

This approach prevents tight coupling between protocol implementations and application logic, enabling independent development and testing of individual components while maintaining system-wide coordination.

FIGURA
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TRABALHO
DO AKOS

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vir a ter

Communication Manager

The Communication Manager provides protocol-agnostic connection state tracking and high-level abstractions for peer communication. Applications and system components interact with remote nodes using Universally unique identifiers (UUIDs) rather than protocol-specific addresses.

For each connection, the manager:

- Handles connection establishment with remote peers across multiple protocols
- Delivers messages to peers, via unicast, multicast or broadcast
- Tracks node state and connection integrity

This unified abstraction enables higher-layer components to interact with peers without concerning themselves with protocol-specific details.

3.6.2 Discovery & Topology Layer

The Discovery & Topology Layer handles peer discovery, network topology maintenance, and multi-hop routing. These components enable autonomous network formation without centralized coordination.

wishful thinking

Peer Discovery Service

The Peer Discovery Service implements multi-protocol neighbor discovery. Short-range protocols like BLE, ESP-NOW, and ZigBee employ active discovery mechanisms (i.e., advertising, peer lists, beacons), while long-range LoRa communication relies on opportunistic message exchange.

The service maintains a unified peer view by:

- Periodically scanning for neighbor advertisements and beacons across active protocols
- Advertising local device capabilities (supported protocols, computational resources, battery status)
- Aggregating discovery information from multiple protocols into a consistent peer database

Discovery handlers translate protocol-specific discovery information into standardized peer descriptors, enabling the Topology Manager to operate on a unified view regardless of the underlying medium.

Topology Manager

Inspired by HyParView [25], the Topology Manager maintains hybrid partial views for scalability and resilience. Each node keeps a small *active view* of peers with which it maintains active communication links, and a larger *passive view* serving as a pool of potential neighbors.

The active view is managed reactively: nodes are added during join operations and removed upon failure detection. The passive view is maintained through periodic shuffle operations that exchange peer descriptors with neighbors, providing view diversity without global knowledge.

Key operations include:

- Active view maintenance with symmetric link enforcement
- Passive view shuffling for backup peer discovery
- Failure detection through connection timeout and explicit heartbeats
- Automatic peer promotion from passive to active view upon failures

TBD.. not sure
how feasible
it is com lo-
ra/ble/esp
now

Routing Service

The Routing Service enables multi-hop communication when direct peer connectivity is unavailable. Unlike tree-based approaches (Section 2.2.4) that restrict communication paths, the Routing Service exploits the mesh topology maintained by the Topology Manager to find alternative routes across different protocols.

Core functionality includes:

- Multi-hop message forwarding using the active view
- Store-and-forward buffering for partitioned networks
- Path selection considering protocol characteristics and device capabilities
- Opportunistic delivery when connectivity windows become available

When networks partition due to range limitations or failures, the Routing Service maintains local buffers of undeliverable messages. Upon partition reconciliation, buffered messages are opportunistically forwarded to their destinations.

3.6.3 Coordination Layer

The Coordination Layer manages protocol adaptation and time synchronization, enabling coordinated operation without master nodes.

wishful think-
ing too

Protocol Selector

The Protocol Selector implements adaptive protocol selection based on message type, device capabilities, and network conditions. Rather than using fixed protocol assignments, it dynamically chooses communication technologies considering multiple factors:

- **Quality of Service (QoS) requirements:** Latency-sensitive messages favor short-range high-throughput protocols (BLE, ESP-NOW), while delay-tolerant data can use long-range protocols (LoRa)
- **Device capabilities:** Protocol choices are constrained by both sender and receiver capabilities obtained from the Protocol Manager
- **Energy budget:** Low-battery devices prefer energy-efficient protocols, potentially sacrificing throughput for lifetime
- **Protocol availability:** Interference or jamming may render specific protocols unavailable, requiring an alternative selection

The selector tracks per-protocol switching costs to prevent excessive oscillation between technologies, ensuring stable operation while remaining adaptable.

Synchronization Service

Coordinated multi-protocol operation requires time synchronization for channel hopping and scheduled communication windows. The Synchronization Service implements decentralized consensus on clock values through randomized gossip, inspired by Randomized gossip-consensus-based sync (RGCS) [50].

Each node maintains a logical clock composed of rate and offset parameters that transform its hardware clock into synchronized logical time. Synchronization occurs through gossip exchanges using Poisson-distributed intervals to minimize collision probability.

Key aspects include:

- Per-protocol Poisson gossip rates accounting for protocol-specific energy costs and reliability
- Converge-to-max criterion for faster convergence than average-based approaches
- Simultaneous rate and offset compensation
- Multi-protocol operation where sync messages can be sent over any available protocol

Unlike master-based synchronization that creates single points of failure, this gossip-based approach maintains synchronization through distributed local interactions.

explicar melhor what I mean by Poisson-distributed?

3.6.4 Data Management Layer

wishful thinking too²

The Data Management Layer optimizes data transmission through compression and efficient queue management, extending operational lifetime in battery-powered deployments.

Compression Service

The Compression Service reduces transmission overhead through prediction-based data reduction (inspired by Ambrosia [43]) and lightweight compression (inspired by Sprintz [7] and the approach in [35]). It employs window-based forecasting where sensor values are transmitted only when prediction error exceeds application-specific thresholds.

Both sender and receiver maintain synchronized prediction state, ensuring consistency without additional communication overhead. The service supports:

- Prediction-based reduction with configurable error thresholds (δ)
- Protocol-aware threshold adaptation (higher thresholds for energy-expensive protocols)
- Lightweight time-series compression using delta coding, zigzag encoding, and Run-length encoding (RLE)
- Graceful degradation under low battery conditions

Protocol-specific compression strategies account for varying energy costs: data transmitted over LoRa may employ more aggressive compression than BLE transmissions, balancing compression overhead against transmission savings.

Message Queue Manager (??)

The Message Queue Manager handles priority-based queuing and buffer management for messages awaiting transmission or forwarding. It maintains separate queues for different message priorities (e.g., critical alerts, periodic sensor data, opportunistic synchronization) and enforces memory limits based on device capabilities.

Responsibilities include:

- Priority queuing with configurable policies **TODO: (strict priority, fair queuing, weighted fair queuing)??**
- Store-and-forward buffer management for partitioned networks
- Delivery confirmation tracking and retransmission scheduling
- Message aging awareness

During network partitioning, the queue manager coordinates with the Routing Service to maintain buffered messages until connectivity is restored, subject to available memory constraints.

3.6.5 Component Interactions

Figure ?? illustrates the interactions between components.

INTERACTION
DIAGRAM

The Protocol Manager provides capability information to the Protocol Selector, enabling adaptive selection decisions. Peer Discovery continuously feeds neighbor information to the Topology Manager, which maintains active/passive views and provides topology information to the Routing Service for multi-hop forwarding decisions.

The Synchronization Service coordinates timing across components, particularly ensuring the Protocol Selector does not switch protocols during synchronization exchanges. The Compression Service reduces message size before they reach the Communication Manager for transmission, while the Message Queue Manager buffers messages and coordinates delivery with the Routing Service.

The Event Dispatcher orchestrates all inter-component communication, enabling loosely coupled interactions through event publication and subscription. This decoupling simplifies component development and incremental system evolution.

PROGRESS REPORT AND PLANNING

4.1 Progress Report

- meia página

4.1.1 Use Cases

- to guide development and evaluation...

4.2 Planning

4.2.1 Validation

- make sure it works and has desired properties in isolation...
- reacts quick... works in set of scenarios... tolerates faults
- maybe compare com mqtt

4.2.2 Task Scheduling

Task 1: Core Components

Task 1.1: Protocol Manager

Task 1.2: Event Dispatcher Validation

Task 1.3: Communication Manager

Task 2: Discovery & Topology

Task 2.1: Peer Discovery Service

Task 2.2: Topology Manager

Task 2.3: Routing Service

Task 3: Coordination Layer

Task 3.1: Protocol Selector

Task 3.2: Synchronization Service

Task 4: Data Management

Task 4.1: Compression Service

Task 4.2: Message Queue

Task 5: Integration

Task 6: Data Management

Task 6.1: Use Case Development

Task 6.2: Experimental Setup

Task 6.3: Performance Evaluation

Task 7: Dissertation Work

Task 7.1: Writing

Task 7.2: Preparation for PerCom (2026)

CHAPTER 4. PROGRESS REPORT AND PLANNING

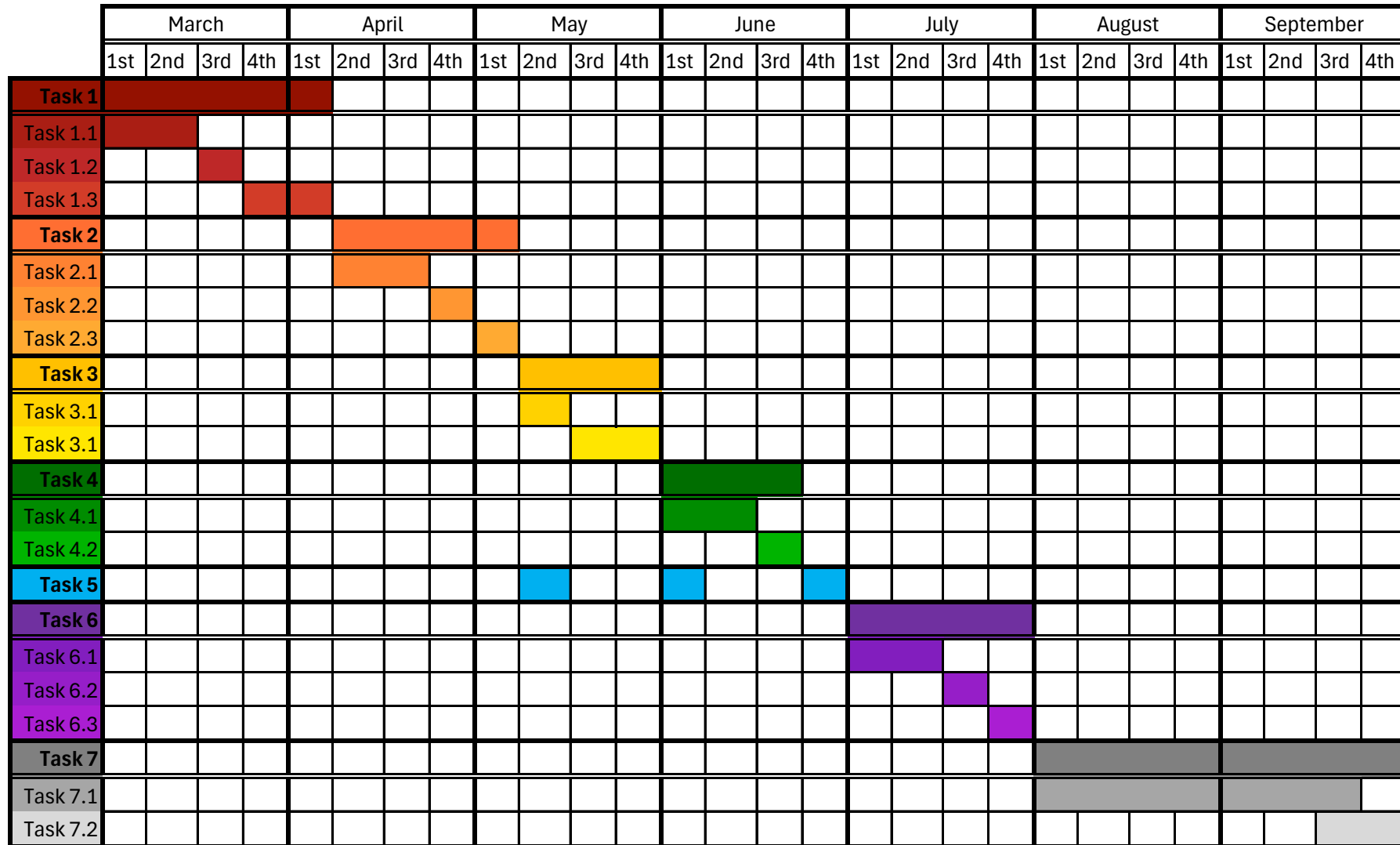


Figure 4.1: GANTT

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