



Solving Interoperability and Performance Challenges over heterogeneous IoT Networks – DNS-based solutions

Antoine BERNARD

November 26th, 2021

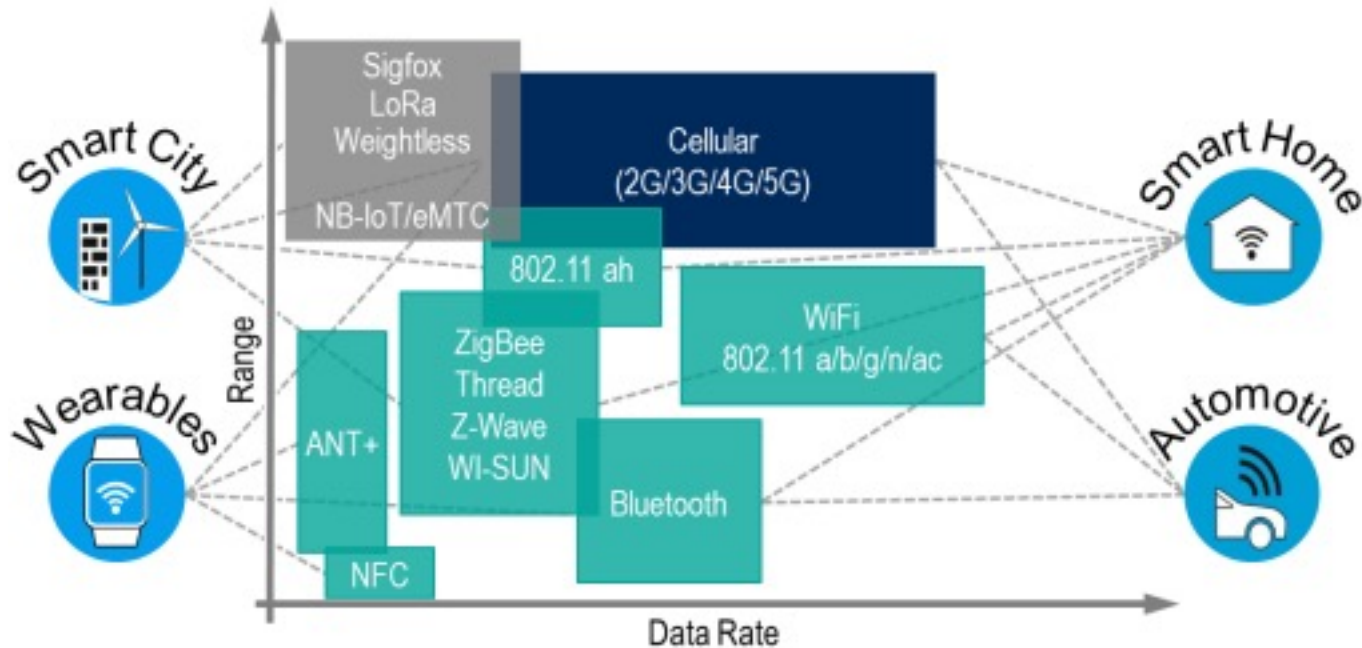
Thesis available at <https://thesis.a-bernard.eu/>

Summary

- **Introduction**
- **Interoperability and device mobility**
 - Building a Roaming Federation
 - Prefetching DNS information
- **Performance challenges**
 - Compressing Headers
 - Minimize network traffic
- **Conclusion**

Introduction

IoT diversity



Source : Whitepaper Rhode & Schwarz

LoRaWAN, a Low-Power Wide-Area Network

■ Constraint network

- Low data rate
- High latency
- Small packet size
- Duty cycle limitations

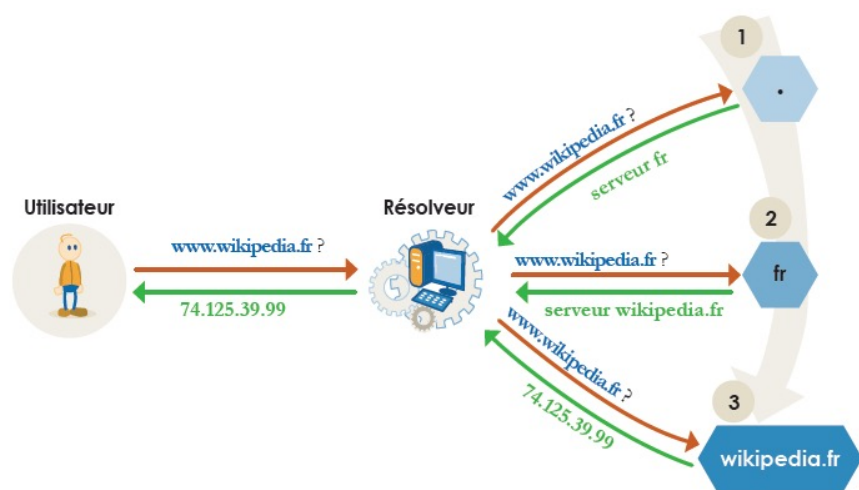
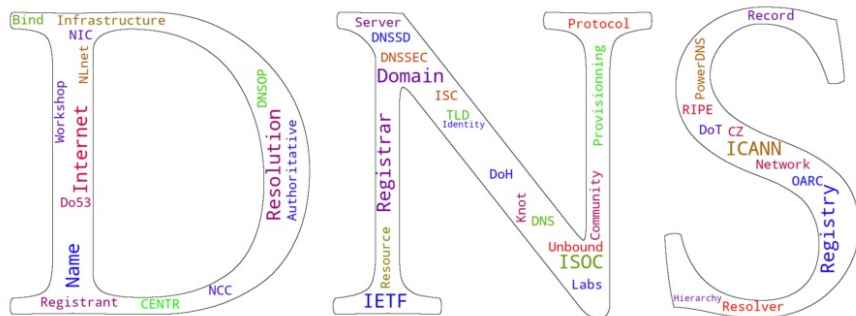


■ Long range

■ Star topology



DNS



Recent evolutions

■ IoT

- Standardization and interoperability are key concerns [1]
- Connect IoT to the Internet [2]
- Adding intelligence to the network [3]

Source :

[1] Debasis Bandyopadhyay and Jaydip Sen. “Internet of Things: Applications and Challenges in Technology and Standardization”. (May 2011)

[2] Michele Zorzi et al. “From today’s INTRANet of things to a future INTERNet of things: a wireless- and mobility-related view” (Dec, 2010)

[3] Benjamin Sliwa, Nico Piatkowski, and Christian Wietfeld. “LIMITS: Lightweight Machine Learning for IoT Systems with Resource Limitations” (June 2020)

■ DNS [4] :

- Over 300 Related RFCs
- 50+ within the last 10 years

■ Key concern :

- Security
 - DoT, DoH
 - DNSSEC
 - DANE
- DNS-SD

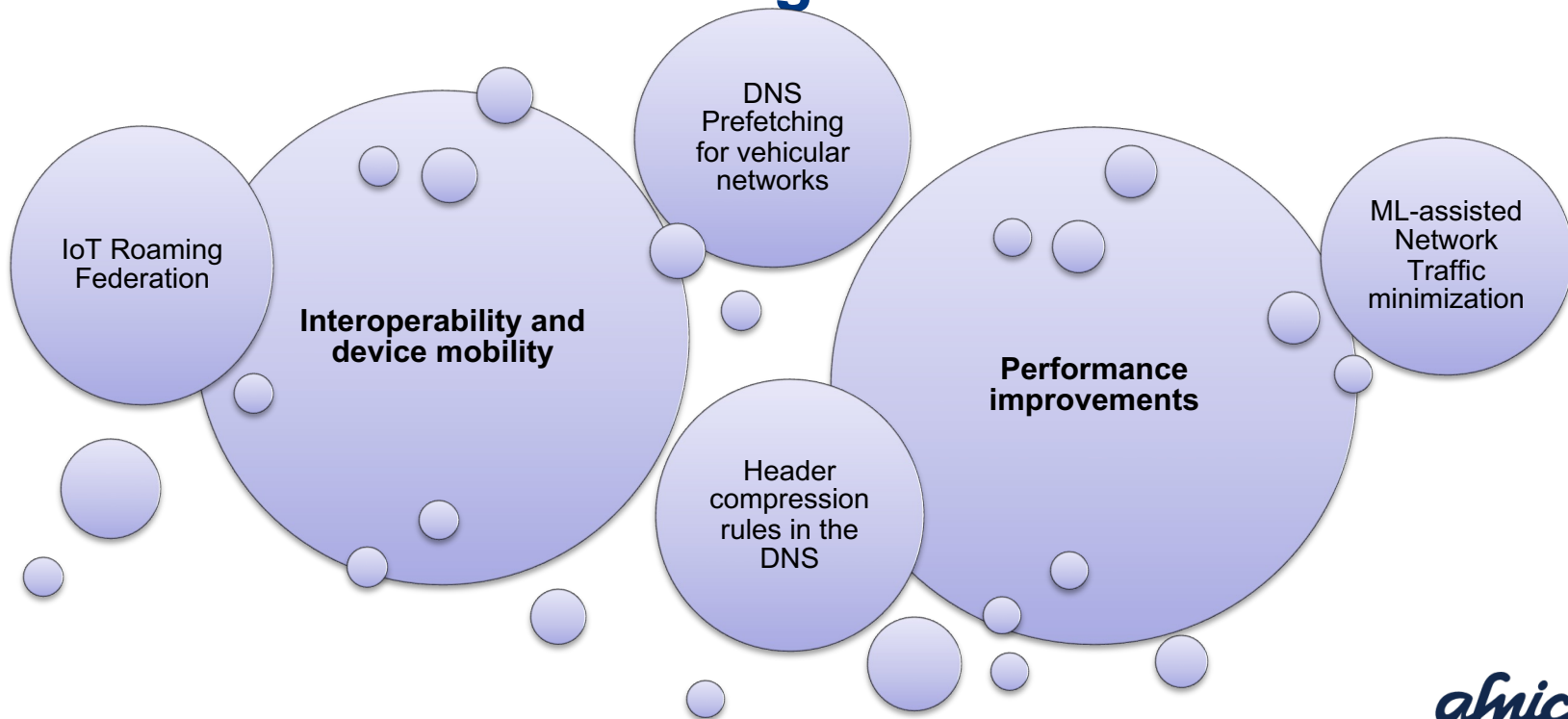
[4] <https://powerdns.org/dns-camel/>



Problem Statement

How can the DNS infrastructure improve IoT architectures and services?

Two main challenges



Improving device mobility

Building an IoT Roaming federation

IoTRoam



Improving mobility ?

- Siloed structures [1]
- Improving mobility
 - Coverage
 - Roaming
- Building Roaming facilitators
 - Peer-to-peer
 - Hub [2]
 - Federation [3]

Reference :

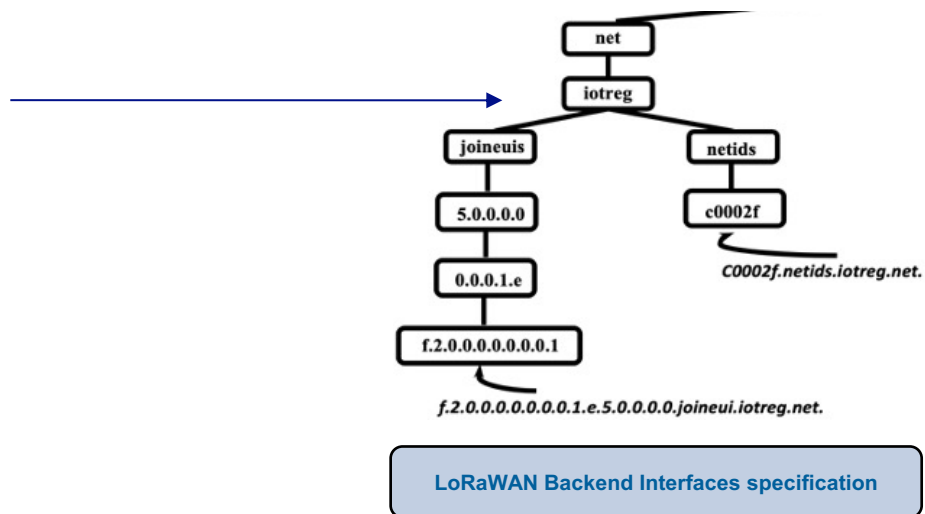
[1] Michele Zorzi et al. “From today’s INTRANet of things to a future INTERNet of things: a wireless- and mobility-related view” (Dec, 2010)

[2] <https://www.thethingsindustries.com/peering/>

[3] <https://eduroam.org/>

Substituting prior configuration

Peer Net-ID
Roaming Policy
Peer's channel plan
Peer's fNS URL
Peer's sNS URL
Peer's NS IP address
Peer JS URL
Peer JS IP Address
Peer JS Http Credentials



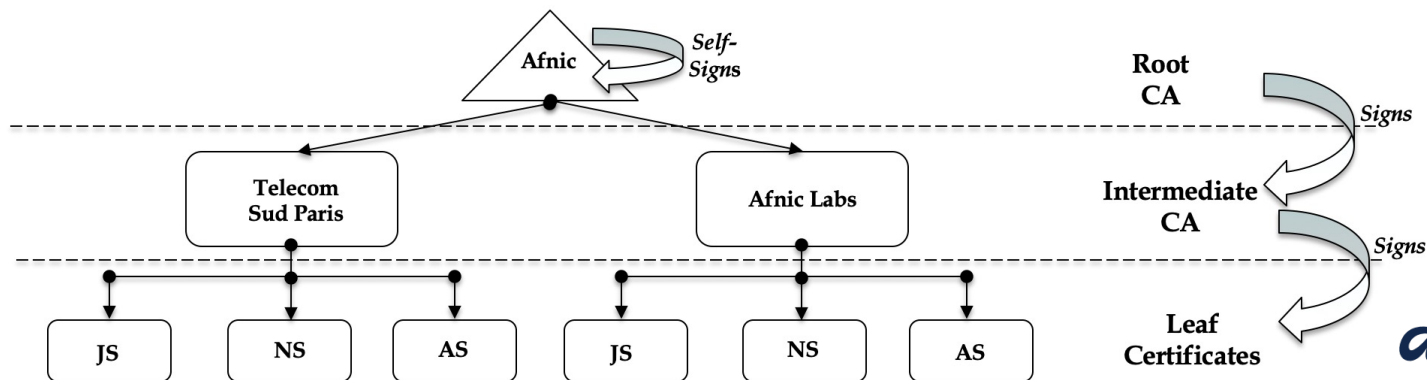
Security concerns

Securing the channel

Building trust between network

Handling global authentication

Conform with LoRaWAN specification

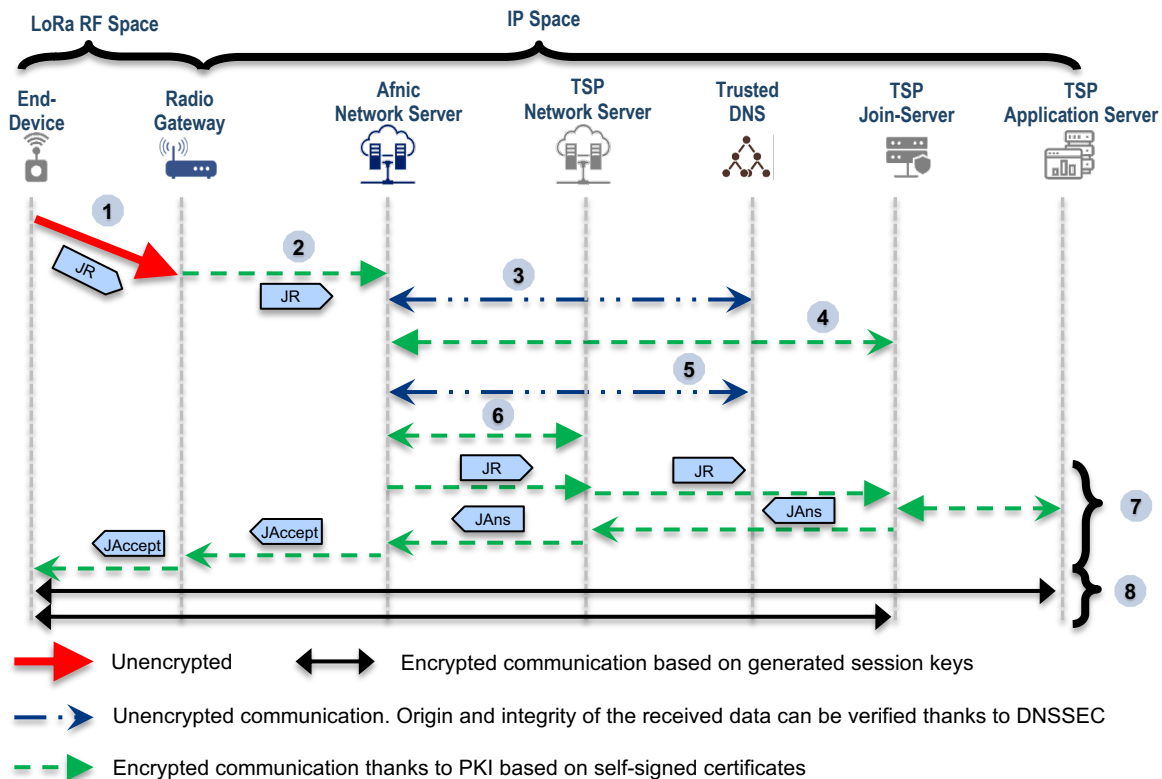


Configuration

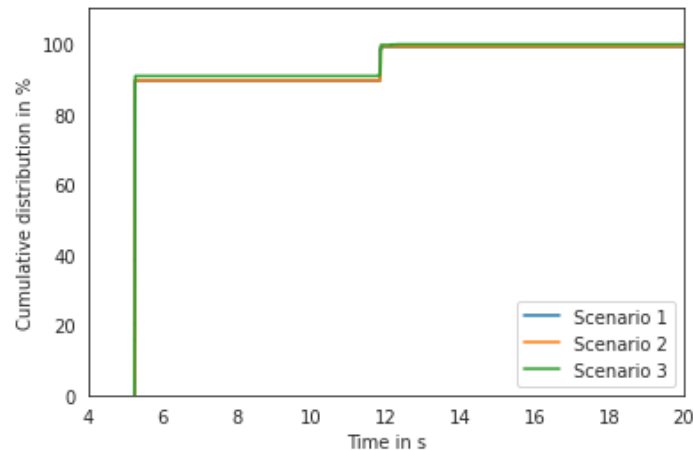
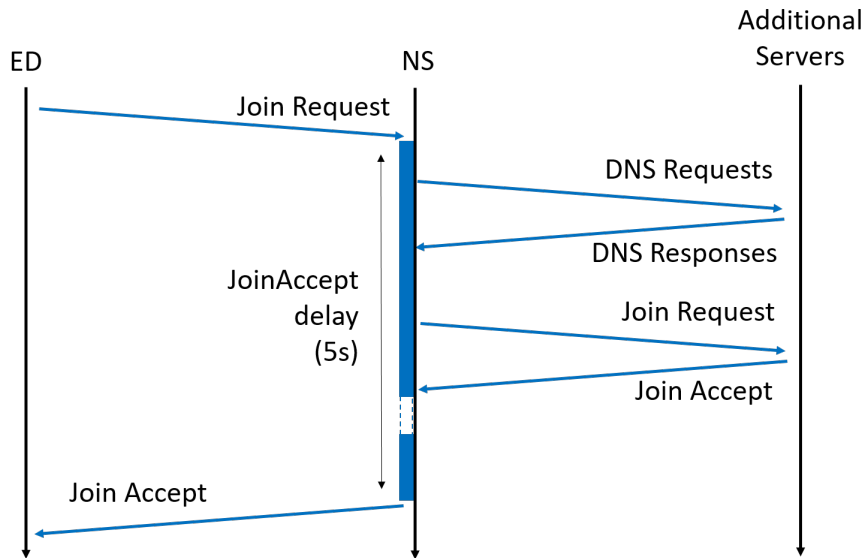
- **DNS (provisionning and autoconfiguration)**
- **Two central ecosystem**
 - DNS Registries
 - Certificate Authorities
- **Use DNS as PKI**
 - Let's encrypt impossible to use
 - Paid certificate make the solution less open
 - DNSSEC and DANE

Revoir

LoRaWAN roaming exchange



Introduced latency ?



- **1. No Roaming**
- **2. Passive Roaming**
- **3. Passive Roaming with additional encryption**

Contributions

- Test infrastructure of LoRaWAN using DNS
- Build and validate the IoT Roam infrastructure
- LoRaWAN specification contribution
- Contribute to the opensource LoRaWAN stack
- Operational, running and open infrastructure

References :

https://lora-alliance.org/resource_hub/lorawan-back-end-interfaces-v1-0/

<https://github.com/brocaar/chirpstack-network-server/releases/tag/v3.11.0>

<https://github.com/AFNIC/IoTRoam-Tutorial/>



Further work

- Explore dual connectivity
- Enhance with new DNS standards:
 - DNSSEC
 - DANE
- Develop the IoTRoam network

Building a roaming federation



afnic
Internet
made in France



Improving device mobility

Prefetching IoT information onto antennas using DNS

Proposal

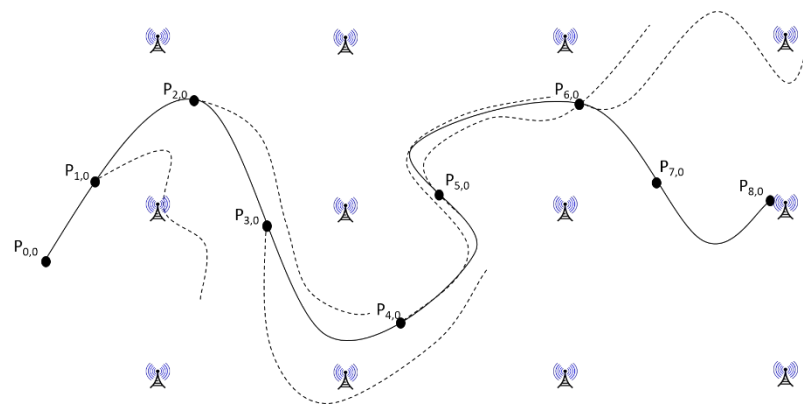
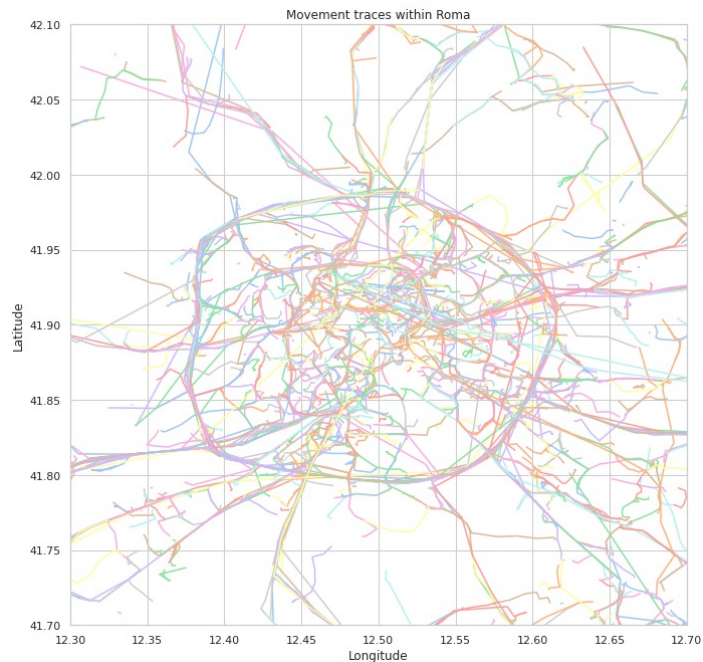
- **Extend uses from the web**
- **Exploit vehicle traffic prediction to realize the prefetching operation**
- **Prefetch using DNS for its high availability**

References :

Driving path stability in VANETs, Laroui et al., 2018

<https://www.chromium.org/developers/design-documents/dns-prefetching>

Methodology - Traces



Methodology – Scenarii

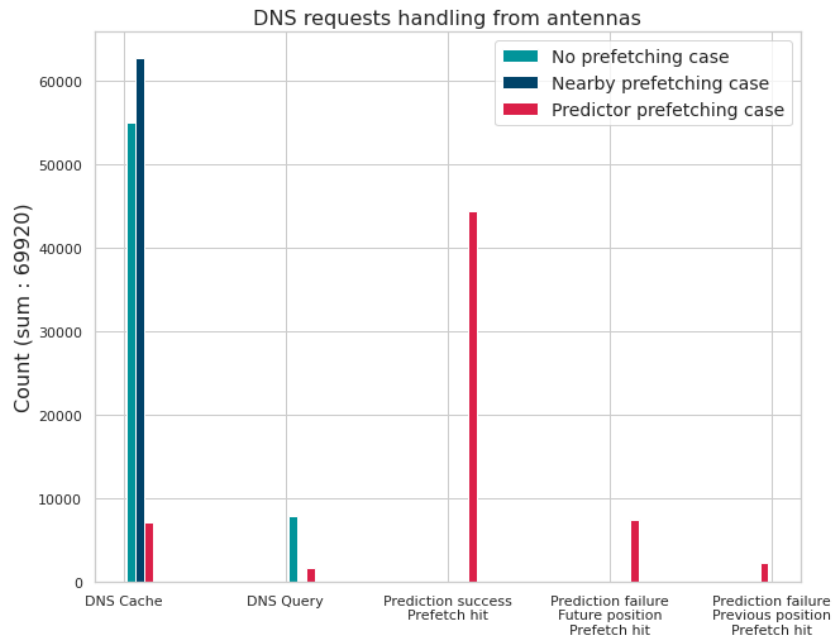
■ 3 scenarii

- No DNS Prefetching
- DNS prefetching on nearby gateway
- DNS prefetching using ML predictor

	Actual position	T+1 Prediction	T+2 Prediction	T+3 Prediction	T+4 Prediction
Antenna ID (T-5)	Z	L	M	N	O
Antenna ID (T-4)	Y	I	J	K	E
Antenna ID (T-3)	X	G	H	D	S
Antenna ID (T-2)	W	F	C	Q	T
Antenna ID (T-1)	V	B	P	R	U
Antenna ID (T)	A				

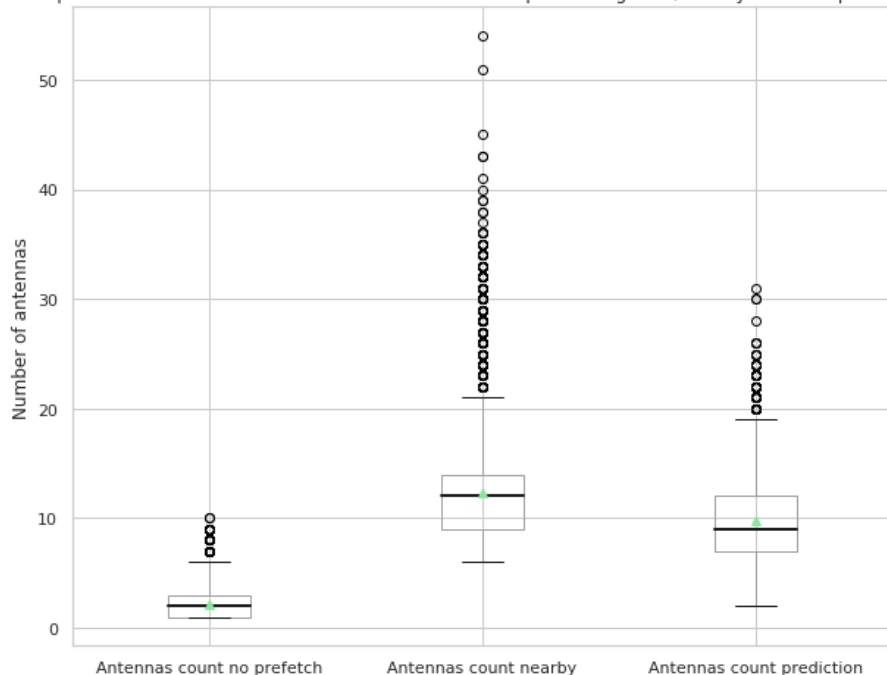
Results

- **80% On-the-fly DNS query reduction**
- **Over 60% exact antenna prediction success**
- **Over 80% prefetched cache hit**
- **Can save up to a second, around 20% of the time allocated in a join procedure**



Results

Comparison of number of solicited antennas between no prefetching case, nearby case and predict case



- **18% less antennas between scenari 2 and 3**
- **Additional study on outliers might be of interest**



Contributions

- **Overall system specification**
- **Prefetching mechanism simulations based on traffic predictor**
- **Caching heating scenario study**
- **Antennas solicitation breakdown**

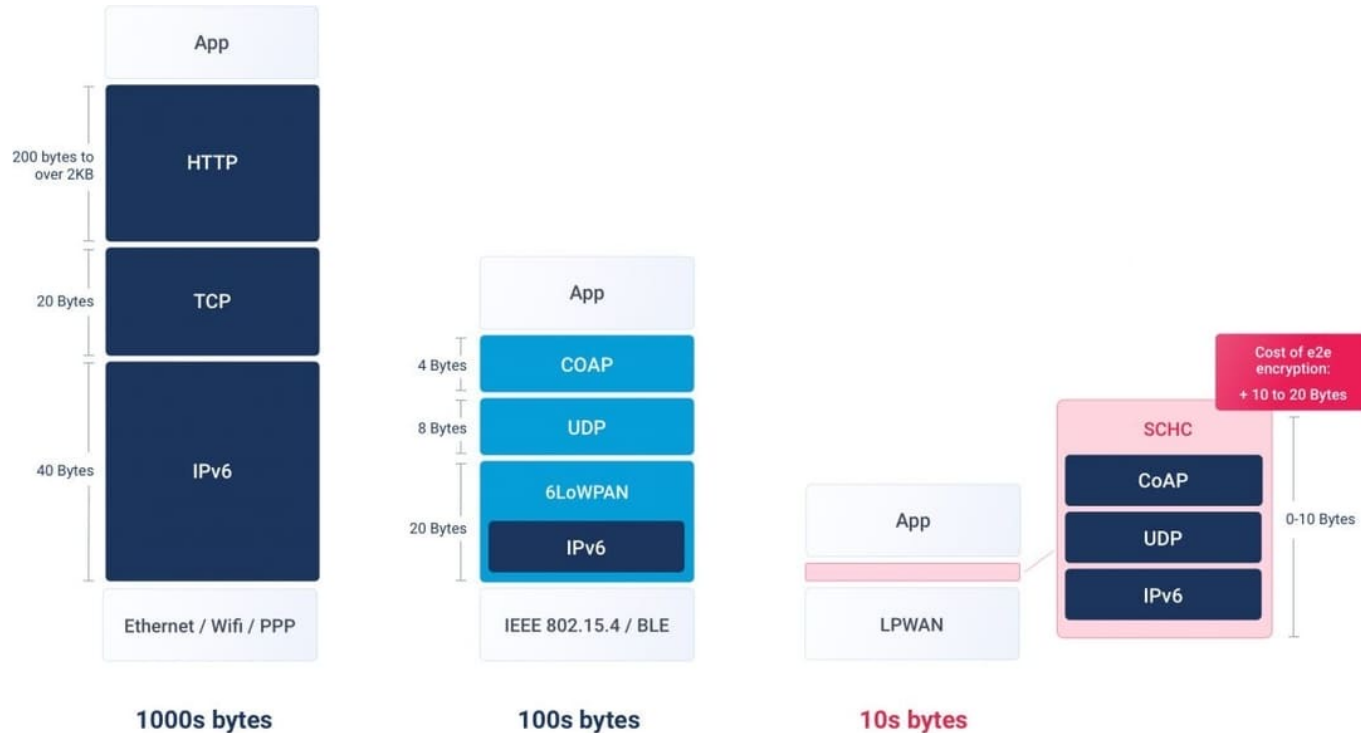
Further work

- Enhance with other traces
- Study other antennas placements
- Predictor quality impact
- More specific division between road topologies
- Impact from traffic density estimations

Solve performance challenges

Compressing Headers

Why compress headers ?



SCHC rule example - extract

```
{  
  "FID": "IPV6.DEV_PREFIX",  
  "FL": 64,  
  "FP": 1,  
  "DI": "Bi",  
  "TV": [ "2001:db8::/64", "fe80::/64", "2001:0420:c0dc:1002::/64" ],  
  "MO": "match-mapping",  
  "CDA": "mapping-sent",  
  "SB": 1  
},  
{  
  "FID": "IPV6.DEV_IID",  
  "FL": 64,  
  "FP": 1,  
  "DI": "Bi",  
  "TV": ">:::79",  
  "MO": "equal",  
  "CDA": "DEVIID"  
},  
{  
  "FID": "IPV6.APP_PREFIX",  
  "FL": 64,  
  "FP": 1,  
  "DI": "Bi",  
  "TV": [ "2001:db8:1::/64", "fe80::/64", "2404:6800:4004:818::/64" ],  
  "MO": "match-mapping",  
  "CDA": "mapping-sent",  
  "SB": 2  
},
```

Source : <https://github.com/openschc/openschc>

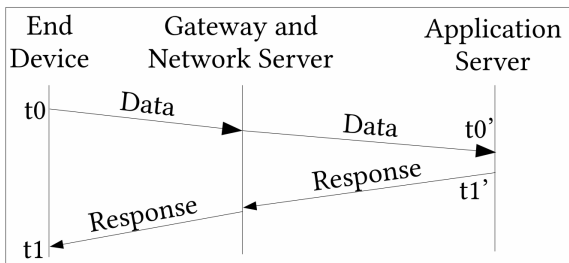
Storing compression parameters

- **Sharing compression rules**
- **Various scenarios studied**
- **Measurements**
 - Decompression Time
 - System latency

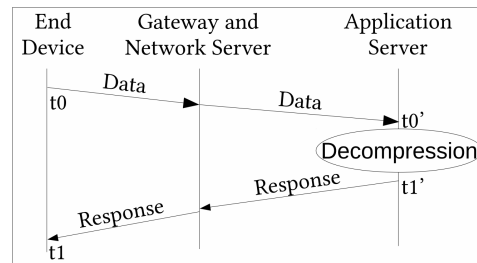
Our proposal

- **Weight constraints would not allow for efficient full rule storage**
- **Store a signature information within the DNS**
- **Mutualize rules when signatures are identical**
- **Fallback onto a web API to get actual rules**

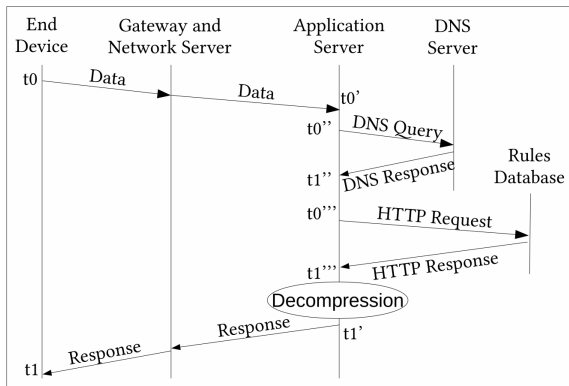
Exchanges experiments



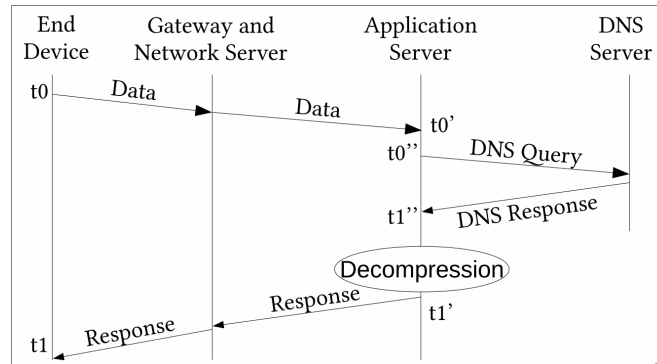
1st Experiment



2nd Experiment

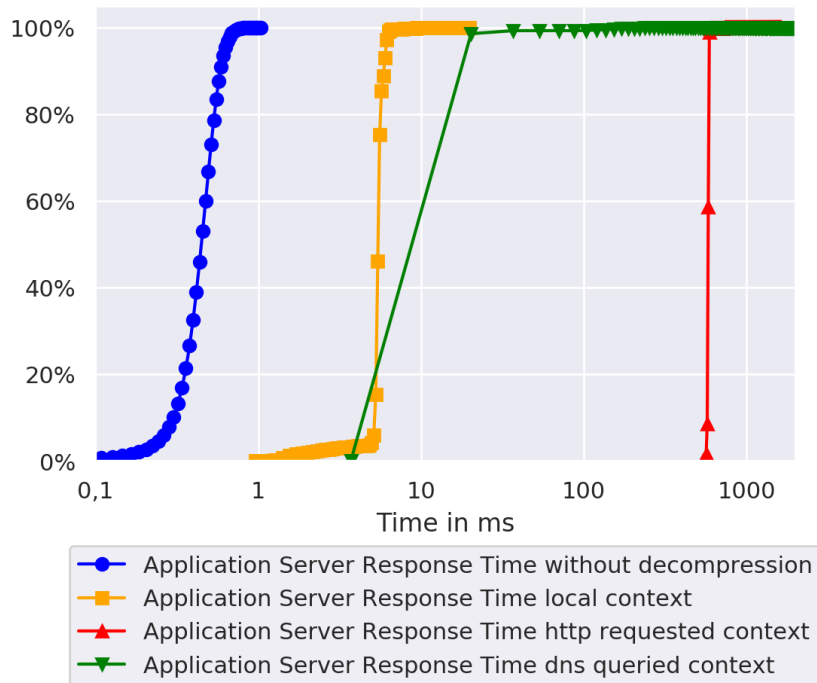


3rd Experiment



4th Experiment

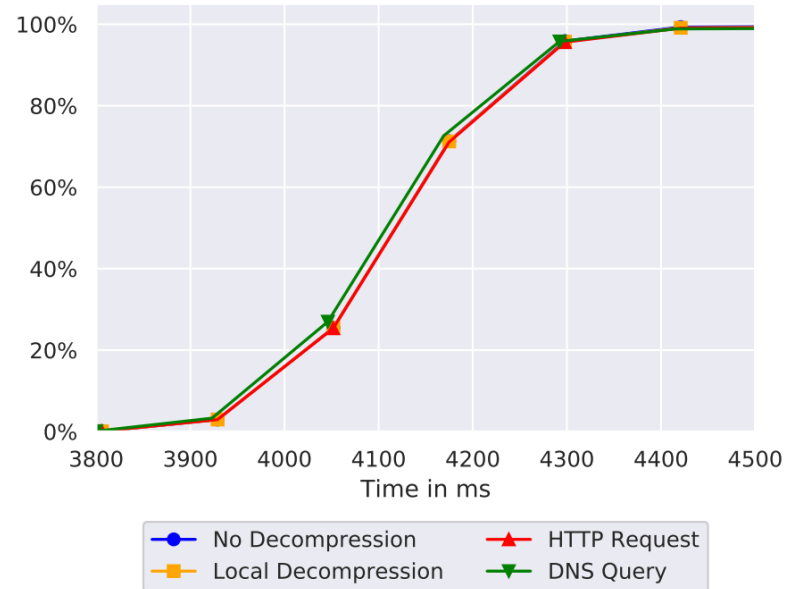
Our results – Decompression Time



- Adding SCHC increase packet processing time up to 8ms
- Using DNS to query the context would take up to 30ms
- HTTP requests are much slower (around 550ms)

Our results – Global Round Trip Time

- No incidence on the communication as the limiting factor is the reception window.
- All responses are received within the same reception window on the device.





Contribution

- **SCHC decompression measurement delays**
- **DNS use for rules querying**
- **Impact from DNS querying using Atlas probes**

Ideas for further work

- **Data Model for Static Context Header Compression**
 - Full rule storage within the DNS ?
- **Discuss our work with the SCHC community at the IETF**

References : <https://datatracker.ietf.org/doc/html/draft-ietf-lpwan-schc-yang-data-model-05>



Solve performance challenges

Compressing application payload

Minimizing traffic



Minimize network traffic from LPWAN sensors

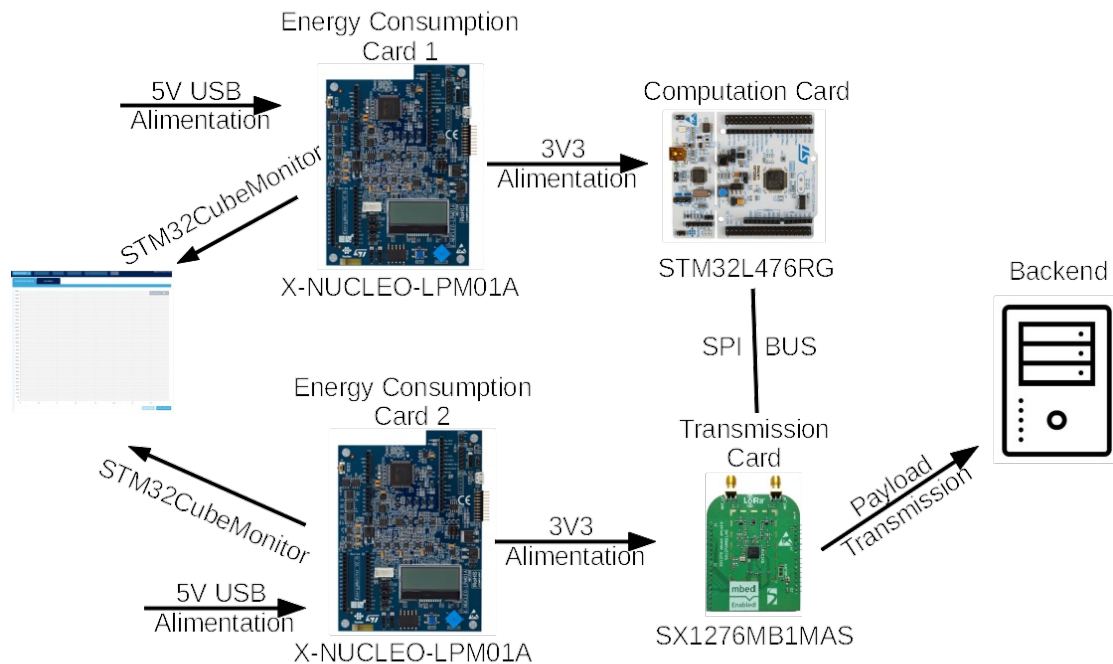
- Scarce resources
- Increase energy efficiency, save battery
- Sensors generate time-correlated data
- Exploit ML techniques to predict this data

Références

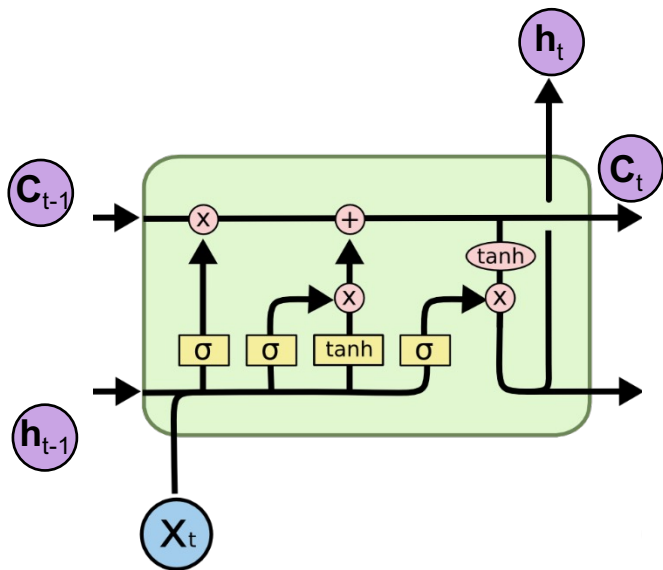
Experimentations

- Tests with various technologies
- Unsupported operations
- Discussions with the TensorFlow Lite community
- Handmade implementation to support the algorithm

Experimental setup



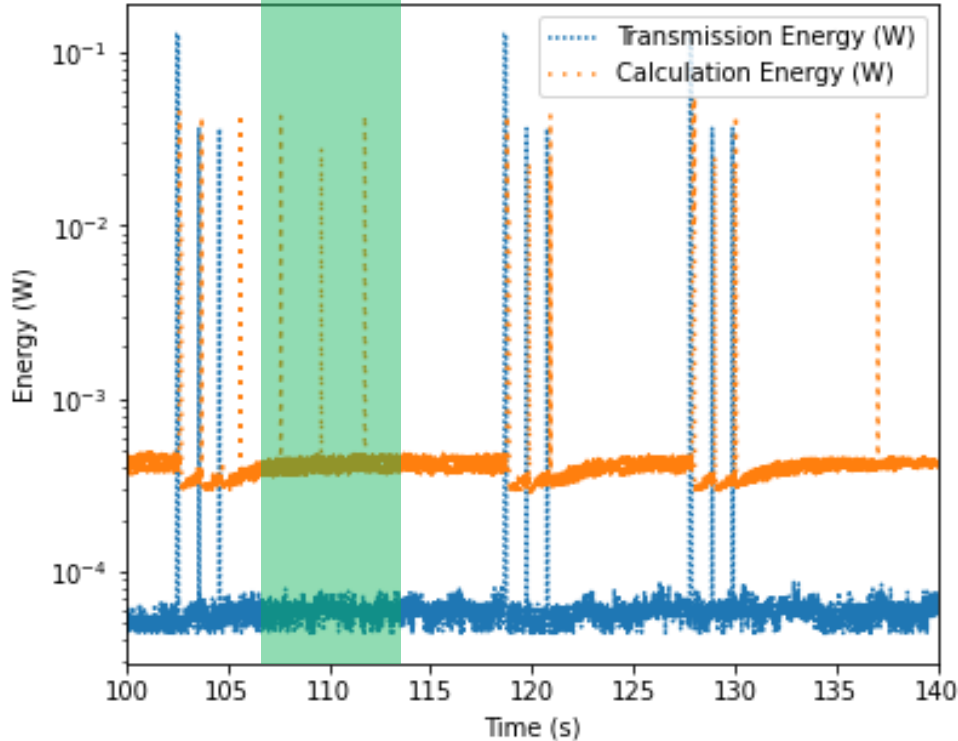
LSTMs and our uses



- LSTM are tools that can correlate Long Term dependancies with Short Term information
- Used in our work to obtain rough time-serie estimation

Source : Christopher Olah, Understanding LSTM Networks, 2015
<https://colah.github.io/posts/2015-08-Understanding-LSTMs/>

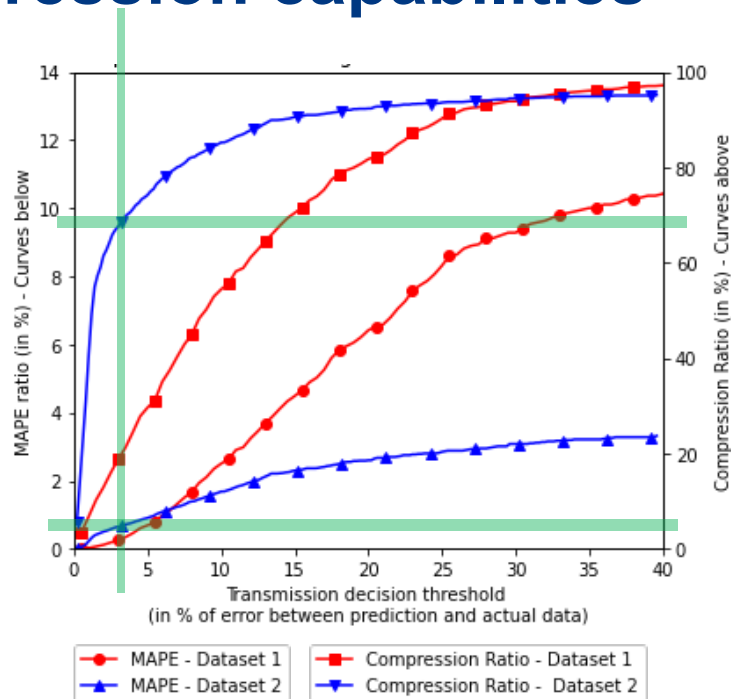
Our Results - Energy



- Two states
- Less activity spikes on the transmission card
- Transmission prevented in the green area
- Energy savings

Our Results – Compression capabilities

- Experimenting with various transmission threshold
- Measuring MAPE & Compression ratio
- Two datasets studied



Results

- Reduction in energy consumed
- Efficient overall with sufficient training data
- **No significance** from the number of cells in neural network
- Quantification is efficient and does not hinder the predictor
- Embedding these algorithms by hand is feasible

Key contributions

- **Implementation of LSTM compatible with MBED OS**
- **Energy measurements**
- **Experimental proof of ML compression schemes applied to networks**

Conclusions and Further work

Conclusions

- **Do not underestimate DNS**
 - Efficient
 - Reliable
 - Secure
- **Roaming is possible within a federated architecture**
- **DNS can store protocol parameters**
- **DNS Prefetching works with predictors**
- **Complex Machine Learning algorithms can be implemented on sensors**

Main contributions

- Tests around roaming for LoRaWAN network including a proposition for IoT Roaming Federation
- Design and performance study of a DNS prefetching scheme based on vehicular traffic prediction
- Design and performance evaluation of a traffic minization scheme based on a sensed data predictor.
- Tests and validation of DNS use for SCHC rules resolution
- LSTM implementation on MBED device and rules sharing

Communications

- **Antoine BERNARD, “La découverte de service à l’aide du DNS”, JCSA 2019**
- **Antoine BERNARD, “LoRaWAN Experimentations”, Doctoral student day @ Telecom SudParis 2019**
- **Antoine BERNARD, “LPWANs tools to scale up IoT solutions from Smart Buildings to Smart Cities”, E4C Summer school: From smart buildings to smart cities, July 2021**
- **Antoine BERNARD, “Embedding ML Algorithms onto LoRaWAN Sensors”, LoRa Alliance Academic WG, Oct 2021**

International conferences

- **Antoine Bernard, Sandoché Balakrichenan, Michel Marot, and Benoit Ampeau, « DNS-based dynamic context resolution for SCHC », IEEE ICC 2020**
- **Antoine Bernard, Aicha Dridi, Michel Marot, Hossam Afifi, and Sandoché Balakrichenan, « Embedding ML Algorithms onto LPWAN Sensors for Compressed Communications », IEEE PIMRC 2021**
- **Sandoché Balakrichenan, Antoine Bernard, Michel Marot, and Benoît Ampeau, « loTRoam – Design and implementation of an openLoRaWAN roaming architecture », IEEE Globecom 2021**
- **Antoine Bernard, Mohammed Laroui, Michel Marot, Sandoché Balakrichenan, Hassine Moun gla, Benoit Ampeau, Hossam Afifi and Monique Becker, « Prefetching of mobile devices information - a DNS perspective », IEEE ICC 2022 (Submitted)**

Possible further work based on this thesis

Expand work on DNS storage [1]

Experiment further on LSTM for MBED device, expanding the work on number of layers and quantification

▪

References : <https://datatracker.ietf.org/doc/html/draft-ietf-lpwan-schc-yang-data-model-05>



Questions ?

Thanks for your attention

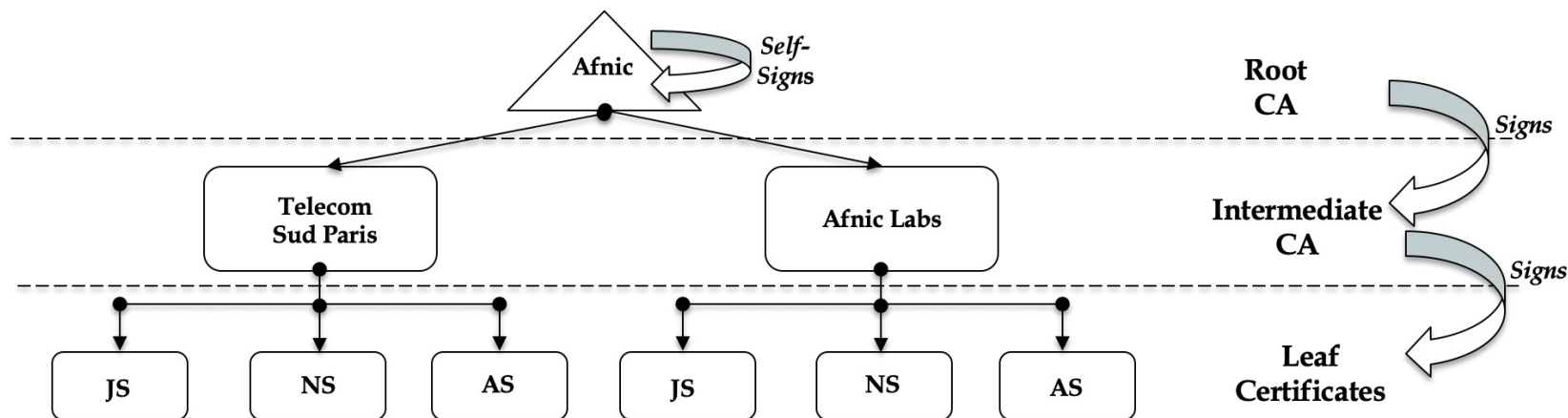
If you need to contact me after this presentation :

thesis@a-bernard.fr

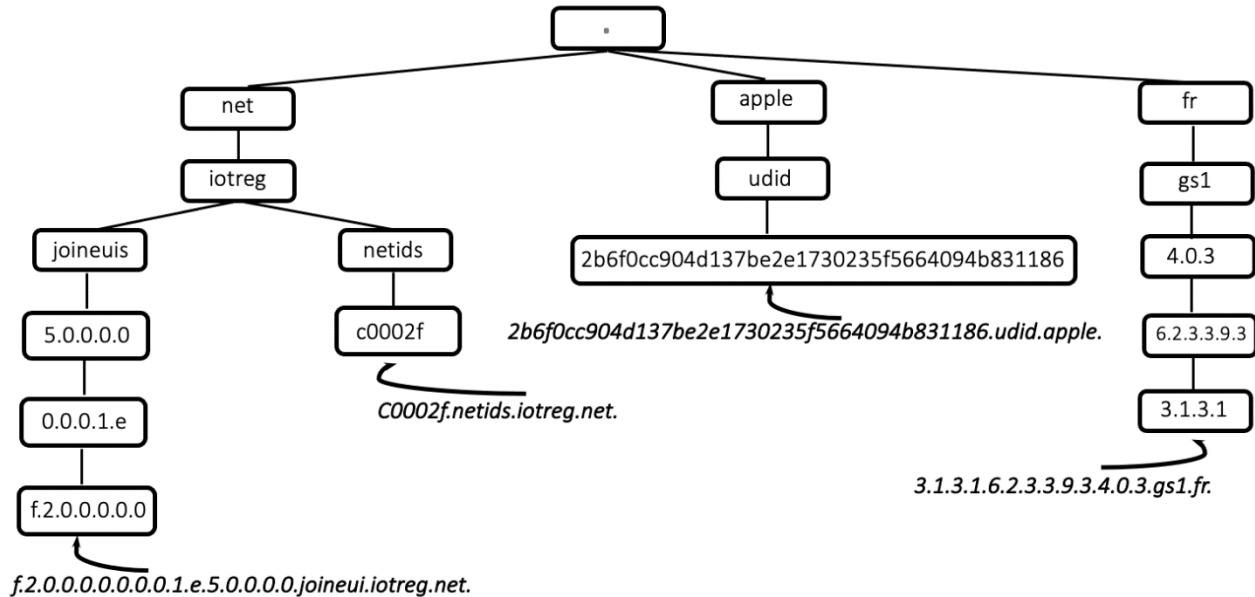
these@a-bernard.fr

Annexes

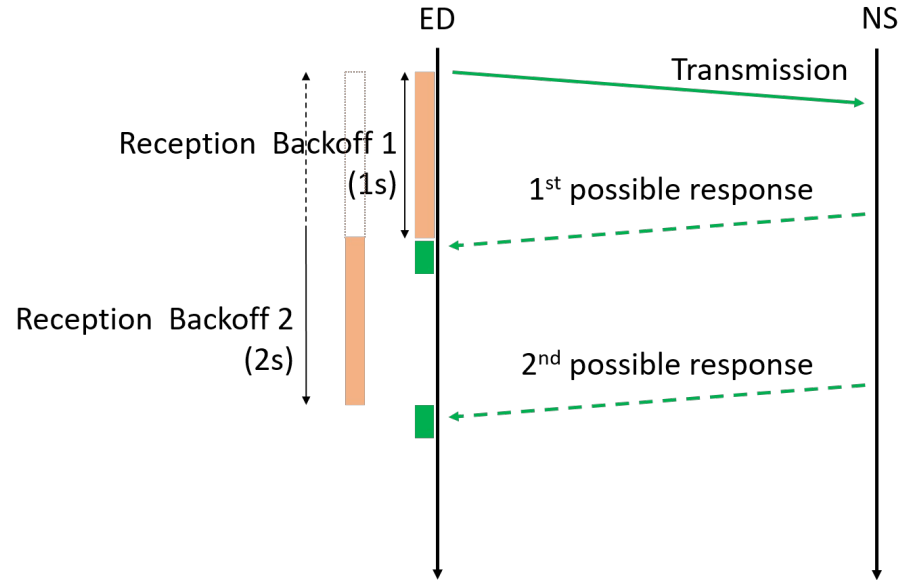
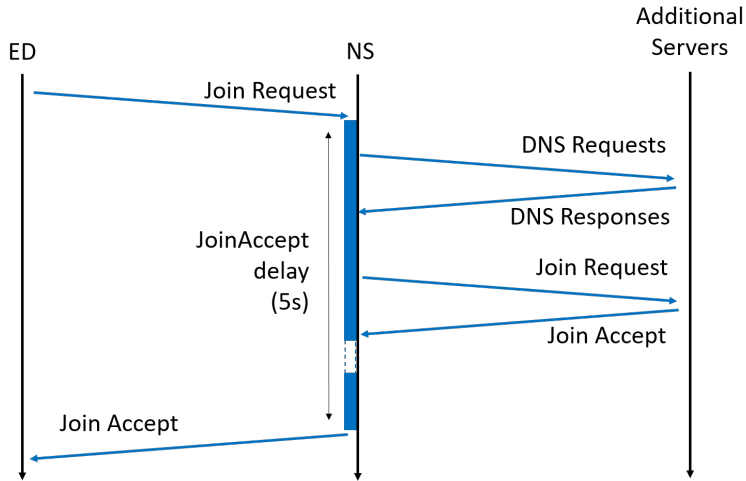
Certificate signing policy



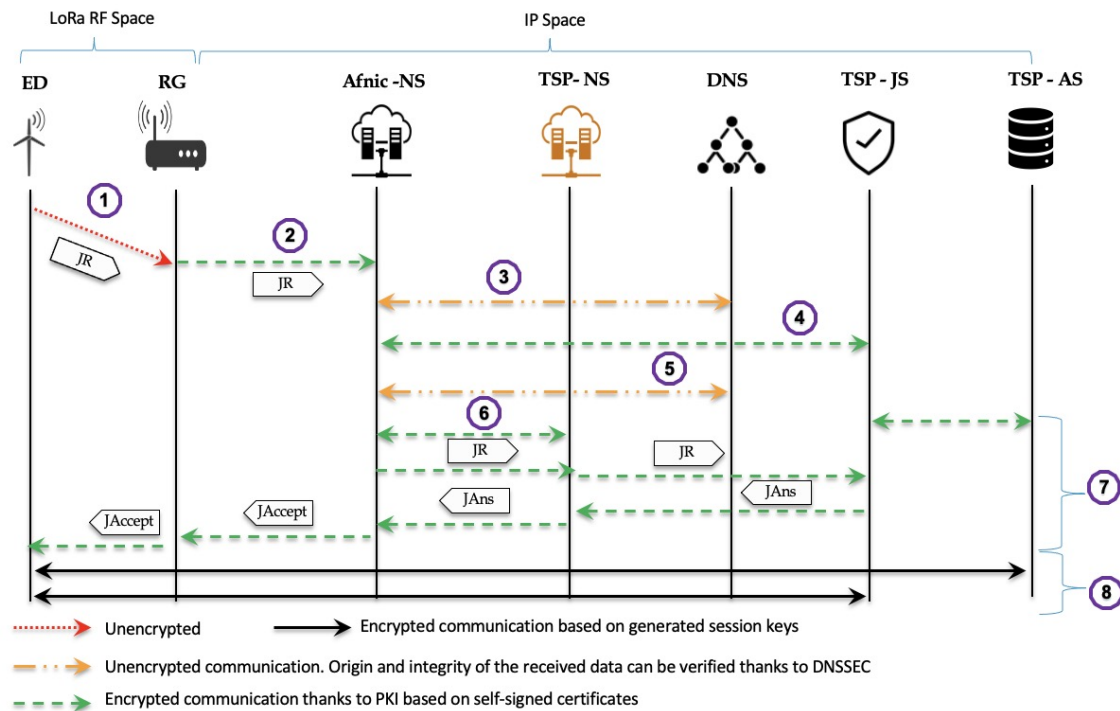
Identifier provisioning through DNS



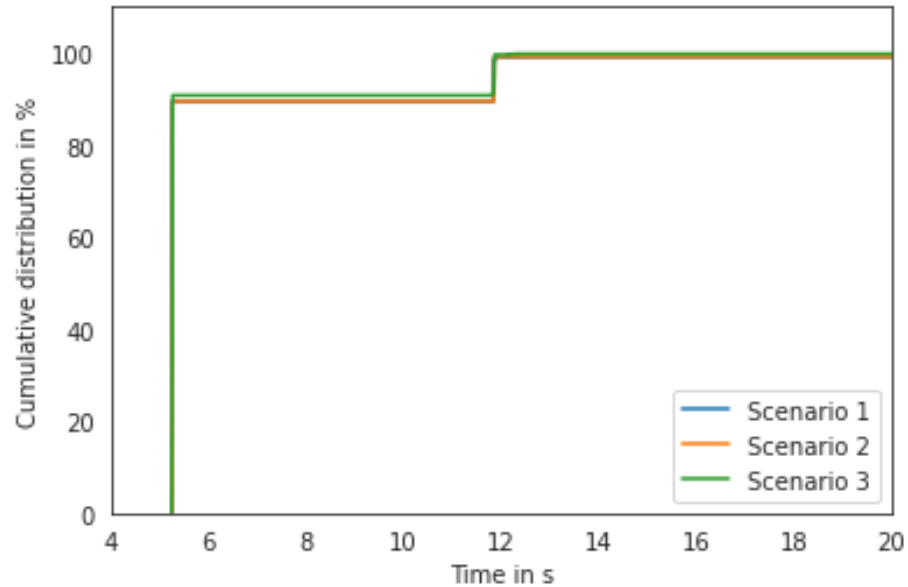
IoT Roam OTAA & Uplink delays



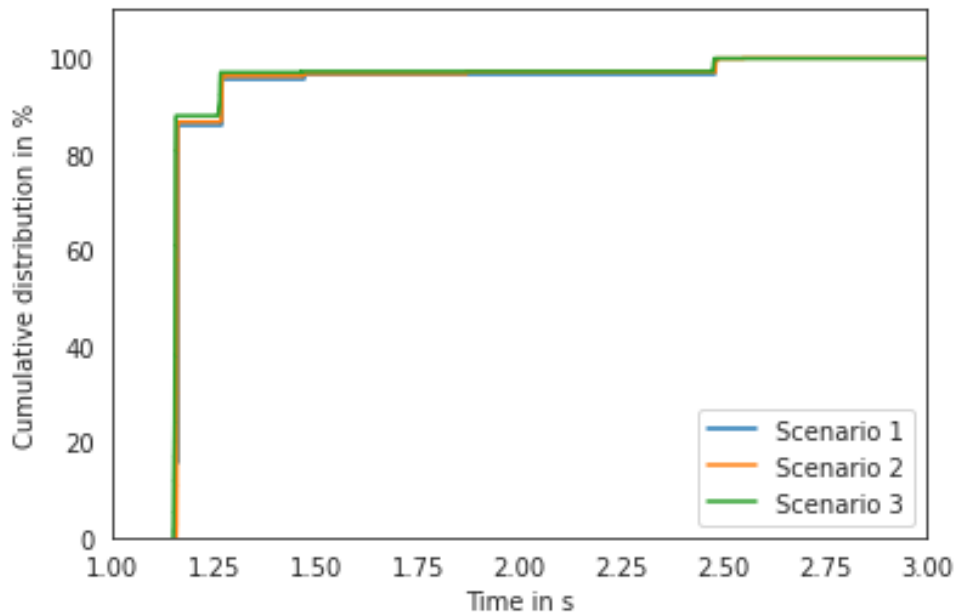
Roaming message flow



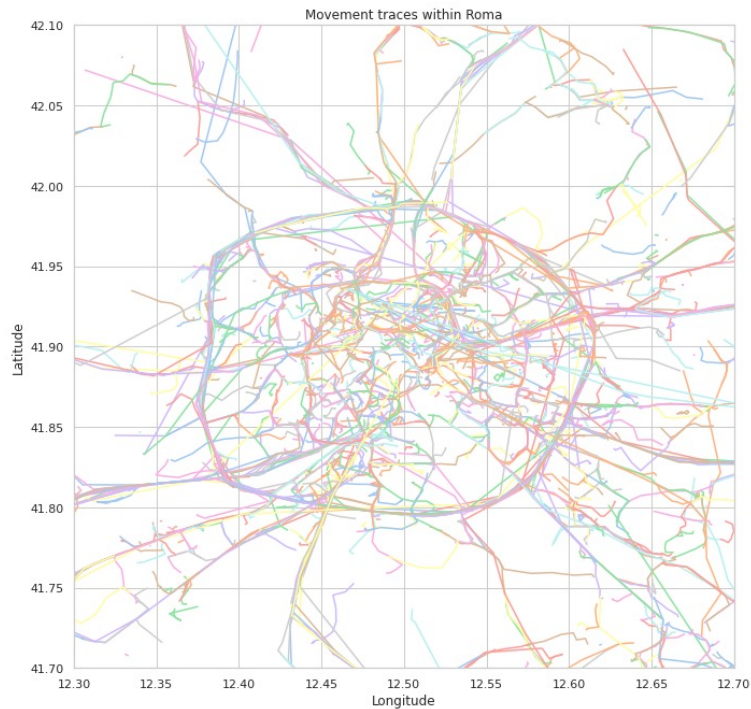
Activation time repartition



First uplink - Ack repartition



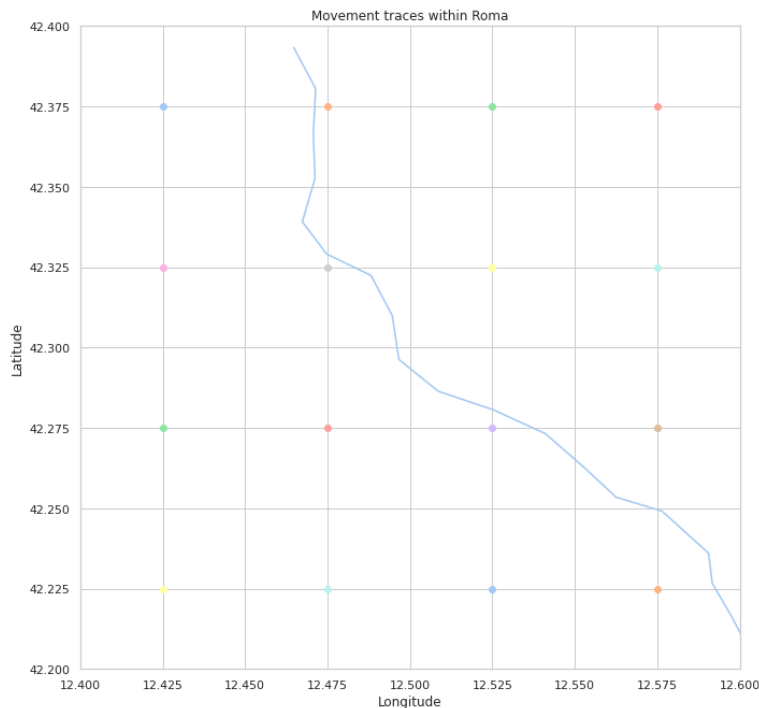
Movement traces



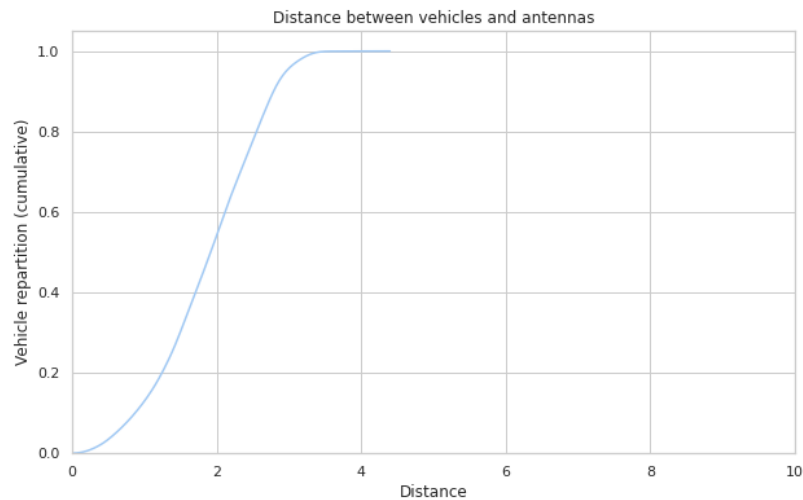
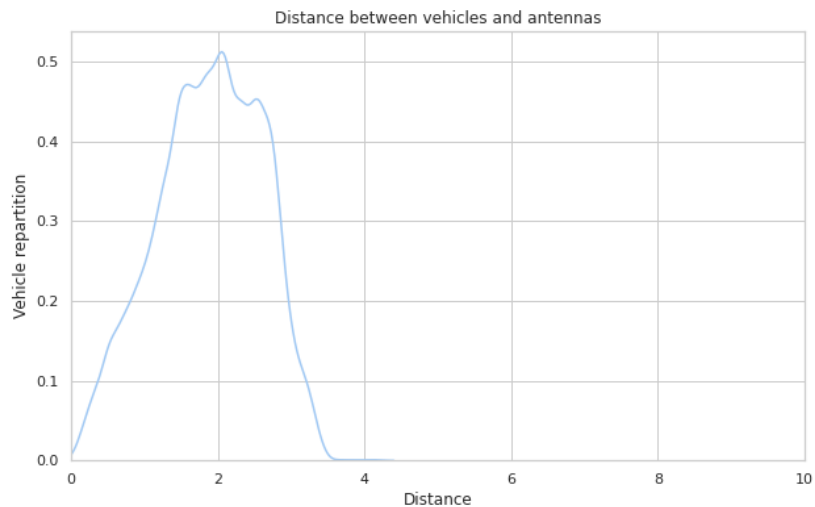
Cache disposition based on predictor

	Actual position	T+1 Prediction	T+2 Prediction	T+3 Prediction	T+4 Prediction
Antenna ID (T-5)	Z	L	M	N	O
Antenna ID (T-4)	Y	I	J	K	E
Antenna ID (T-3)	X	G	H	D	S
Antenna ID (T-2)	W	F	C	Q	T
Antenna ID (T-1)	V	B	P	R	U
Antenna ID (T)	A				

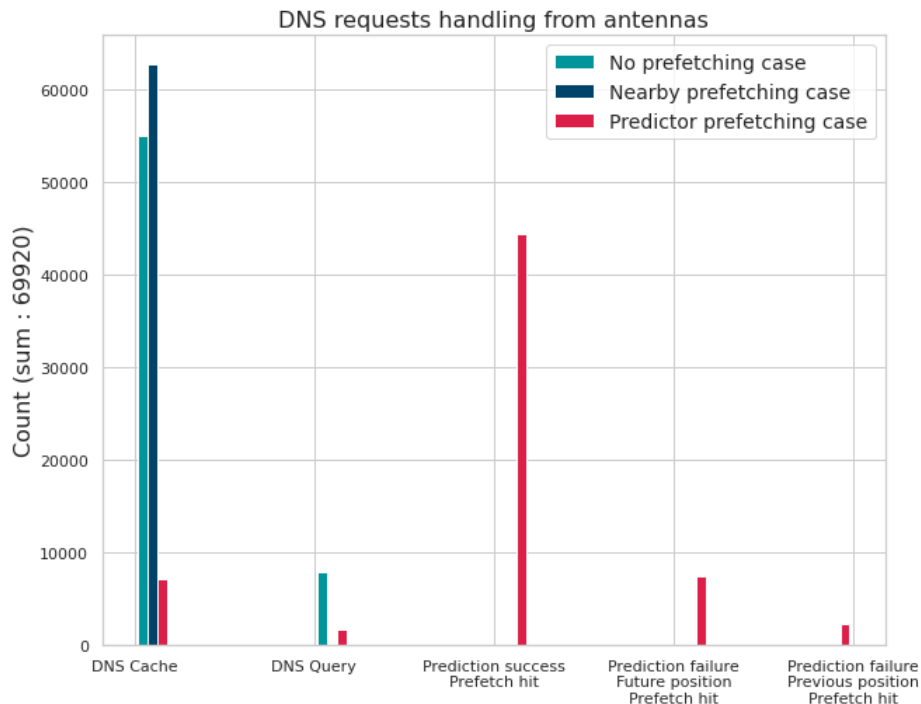
Movement trace and antennas



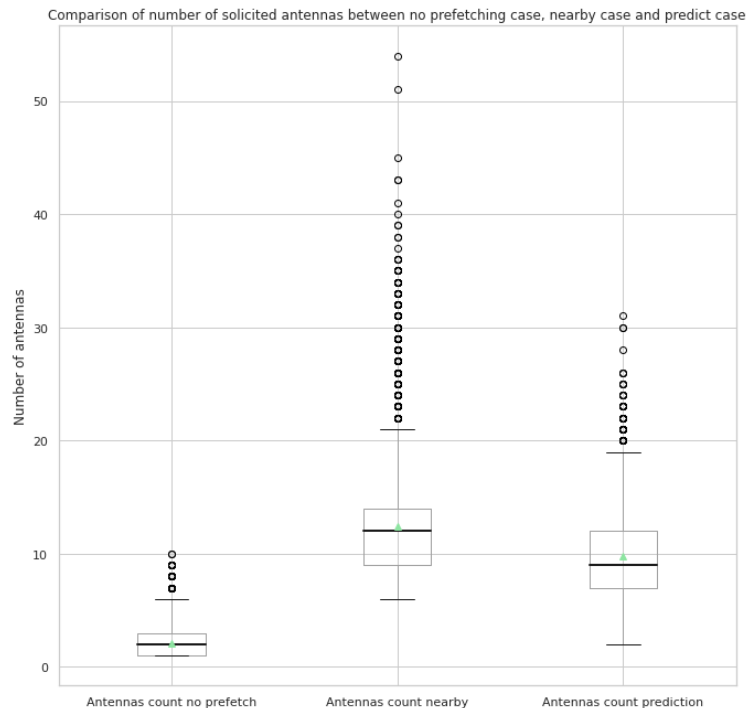
Antenna-vehicule distance



DNS Queries between scenarios



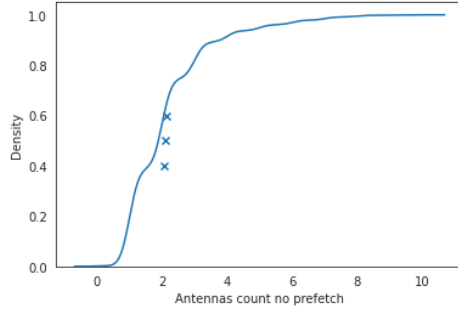
Number of antennas solicited



Antennas solicitation

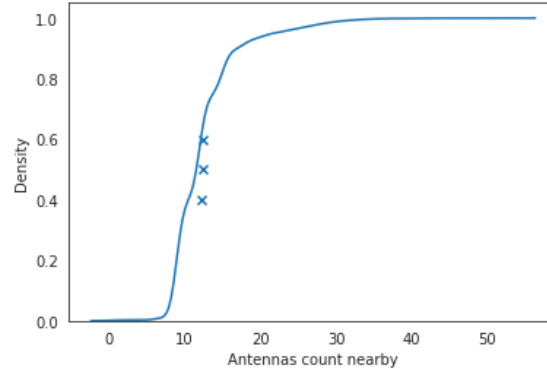
(2.0738689084017174, 2.1058789954337898, 2.137889082465862)

Solicited antennas repartition no prefetching case - Cumulative



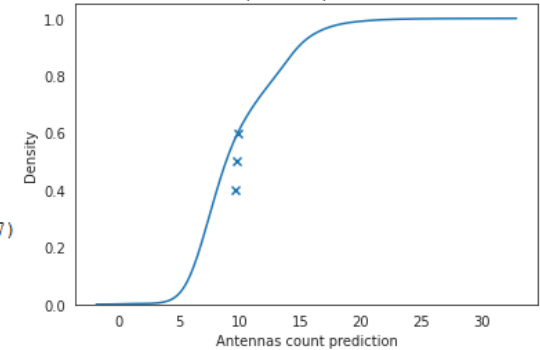
(12.259453278600654, 12.367294520547945, 12.475135762495237)

Solicited antennas repartition nearby case - Cumulative



(9.6714542796611, 9.754994292237443, 9.838534304813786)

Solicited antennas repartition prediction case - Cumulative



Frame size LPWANs

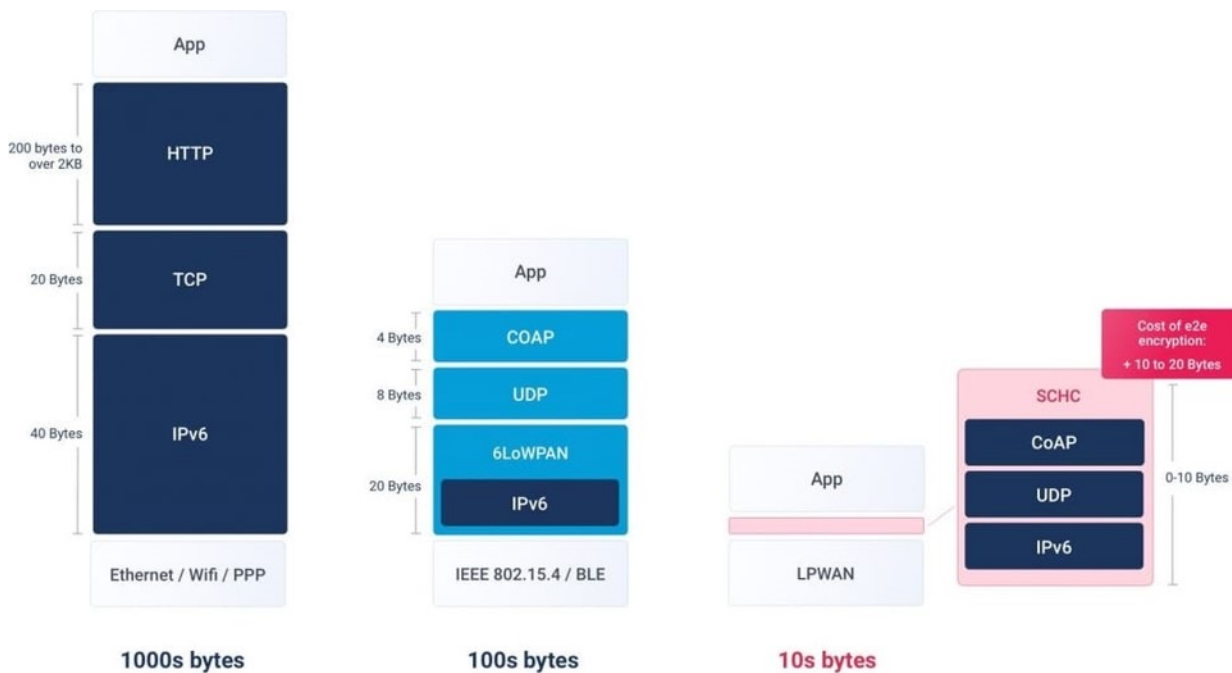
	LoRaWAN (bytes)	NB-IoT/LTE-M (bytes)	SigFox (bytes)
Frame size	250	1600	29

TABLE 4.1: Max Frame size from the main LPWANs technologies

Headers	LoRaWAN	NB-IoT/LTE-M	SigFox
L2 header	8 octets 3.2 %	14 octets 0.875 %	10 octets 34,4 %
L3 / IPv6 header (40 bytes)	16 %	2.5 %	138 %
L4 / UDP header (8 bytes)	3.2 %	.5 %	27.6 %
L5 / CoAP header (4 bytes)	1.6 %	.25 %	13.8 %
L3+L4+L5 / SCHC (2 bytes)	0.8 %	.125 %	6.9 %
Cumulative (no SCHC)	24 %	4.125 %	213.8%
Cumulative (SCHC)	4 %	1 %	41.3 %

TABLE 4.2: Frame Header Occupation as percentage of frame size for the main LPWANs technologies

Explaining SCHC



SCHC rule example

Rule 0

Field	FL	FP	DI	Value	Match Opera.	Comp Decomp Action	Sent [bits]
IPv6 Version	4	1	Bi	6	ignore	not-sent	
IPv6 DiffServ	8	1	Bi	0	equal	not-sent	
IPv6 Flow Label	20	1	Bi	0	equal	not-sent	
IPv6 Length	16	1	Bi		ignore	compute-*	
IPv6 Next Header	8	1	Bi	17	equal	not-sent	
IPv6 Hop Limit	8	1	Bi	255	ignore	not-sent	
IPv6 DevPrefix	64	1	Bi	FE80::/64	equal	not-sent	
IPv6 DevIID	64	1	Bi		ignore	DevIID	
IPv6 AppPrefix	64	1	Bi	FE80::/64	equal	not-sent	
IPv6 AppIID	64	1	Bi	:::1	equal	not-sent	

SCHC measurement platform

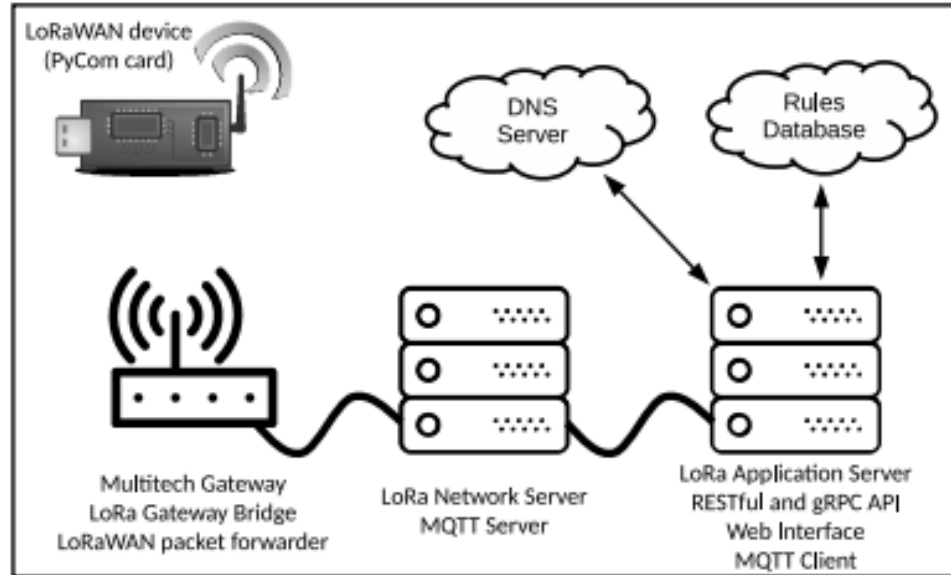
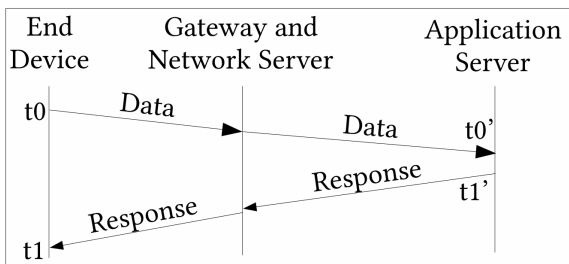
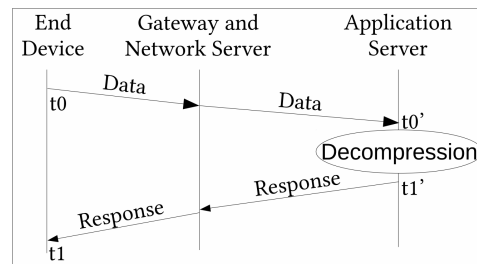


FIGURE 4.3: Measurement Platform's Network and system architecture (rework this scheme)

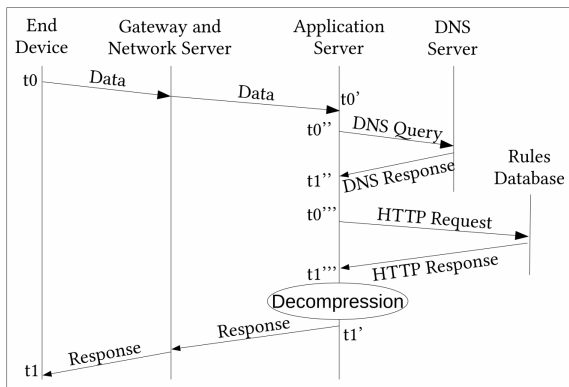
SCHC measurements scenari



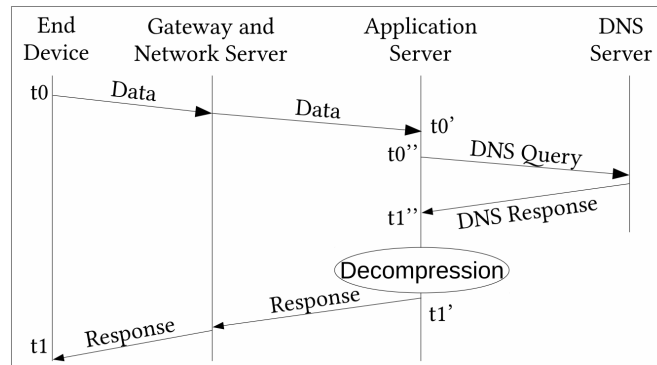
1st Experiment



2nd Experiment



3rd Experiment



4th Experiment

AS response time – decompression impact

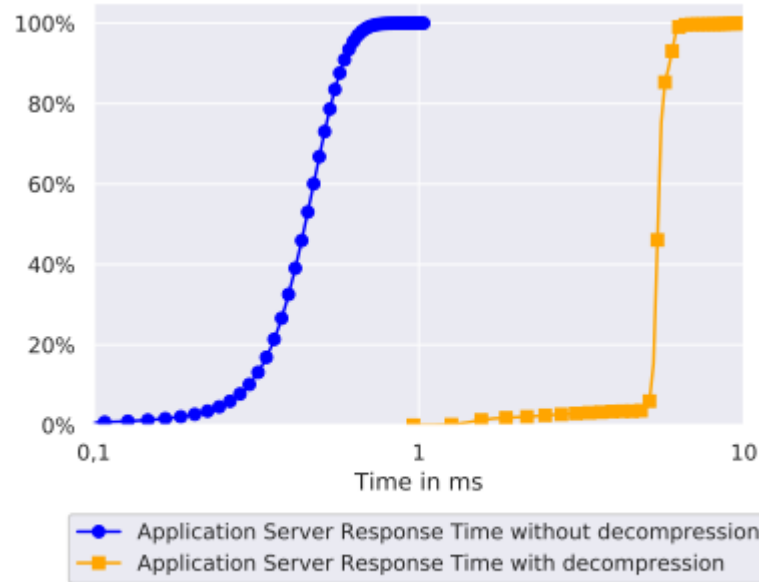


FIGURE 4.8: Cumulative distribution function of the AS Response Time $t1' - t0'$ (in %) against time in ms for Scenarios 1 and 2

AS response time – remote querying impact

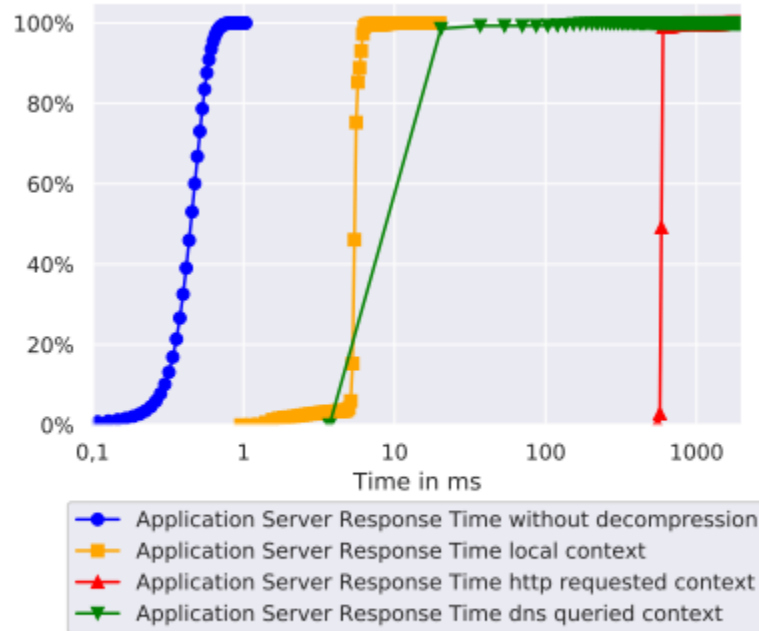


FIGURE 4.9: Cumulative distribution function of the AS Response Time $t1' - t0'$ (in %) against time in ms for all scenarios

DNS Response Time using Atlas probes

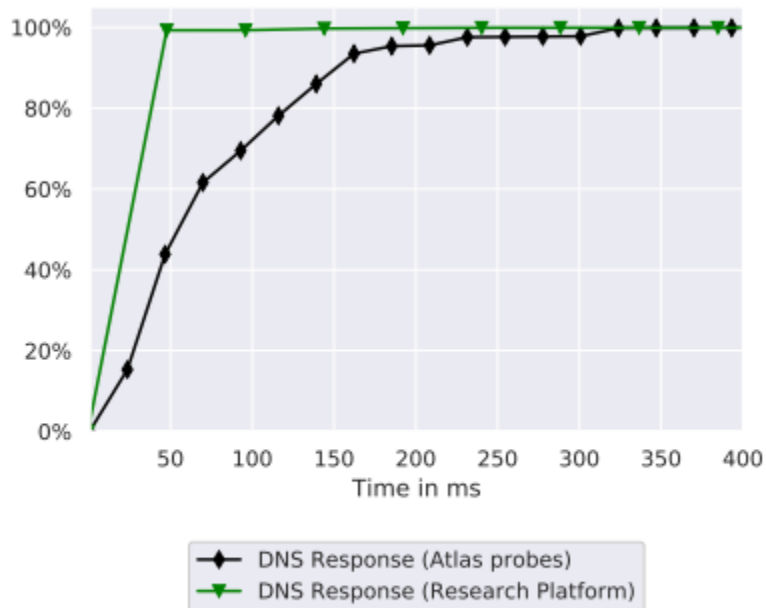


FIGURE 4.10: Cumulative distribution function of the DNS Response Time $t1^{\theta} - t0^{\theta}$ (in %) against time in ms for Scenario 3 compared and from RIPE Atlas [\[327\]](#) Measurements

Global RTT

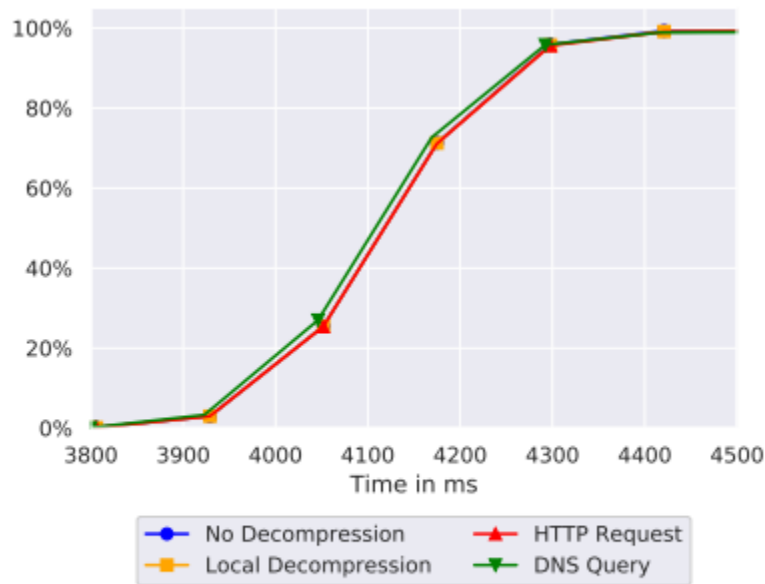
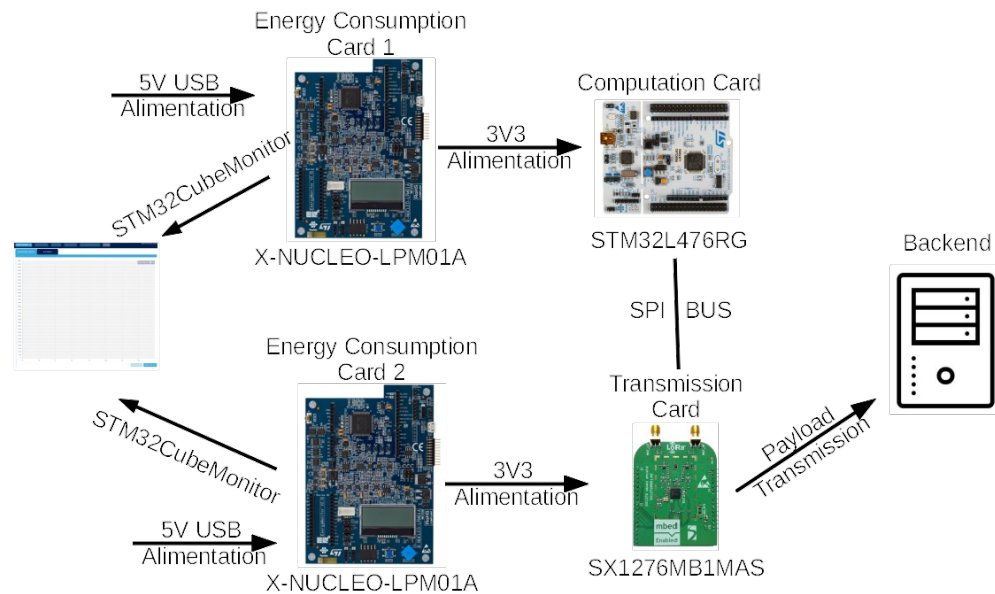
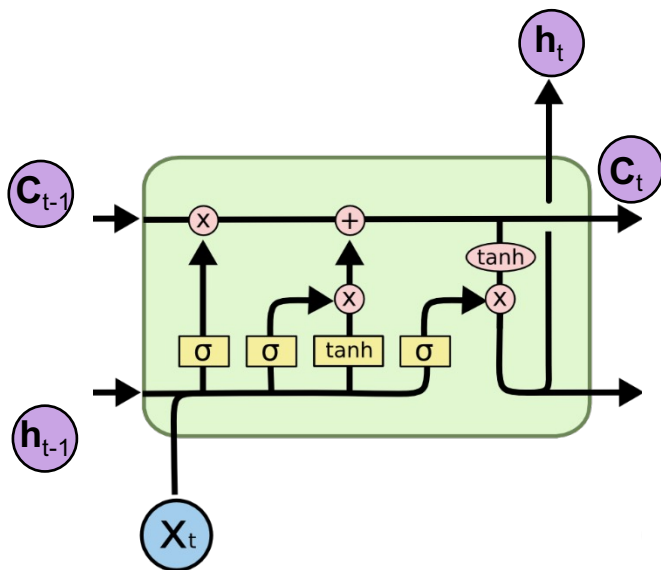


FIGURE 4.11: Cumulative distribution function of the RTT $t_1 - t_0$ (in %) against time in ms for all scenarios (all the curves are the superposed)

ML Energy Measurements platform



LSTM



forget gate :

$$f_t = \sigma(W_f \cdot [h_{t-1}, x_t] + b_f)$$

input gate :

$$i_t = \sigma(W_i \cdot [h_{t-1}, x_t] + b_i)$$

candidate value:

$$\tilde{C}_t = \tanh(W_C \cdot [h_{t-1}, x_t] + b_C)$$

new cell state:

$$C_t = f_t * C_{t-1} + i_t * \tilde{C}_t$$

output:

$$h_t = \sigma(W_o \cdot [h_{t-1}, x_t] + b_o) * \tanh(C_t)$$

Christopher Olah, Understanding LSTM Networks, 2015

<https://colah.github.io/posts/2015-08-Understanding-LSTMs/>

Energy consumption

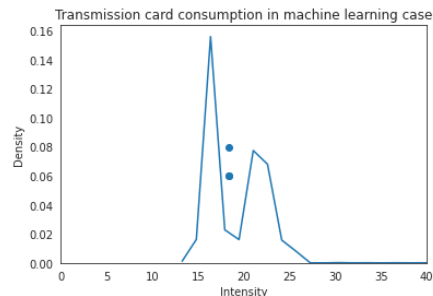
TABLE 5.1: Comparison of the mean energy consumption of the calculation card and its variance, with and without LSTM-based compression (in Watts)

With Machine Learning		Without Machine Learning	
Mean value (W)	Variance	Mean value (W)	Variance
$6.31 * 10^{-4}$	$7.57 * 10^{-5}$	$7.76 * 10^{-4}$	$7.61 * 10^{-5}$

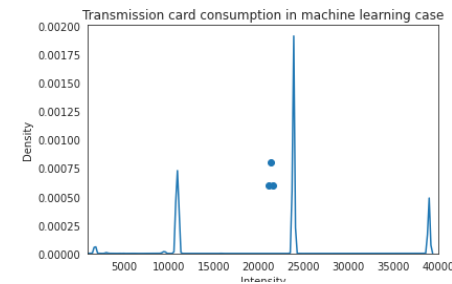
TABLE 5.2: Comparison of the mean energy consumption of the transmission card and its variance, with and without LSTM-based compression (in Watts)

With Machine Learning		Without Machine Learning	
Mean value (W)	Variance	Mean value (W)	Variance
$5.48 * 10^{-4}$	$4.10 * 10^{-5}$	$9.87 * 10^{-4}$	$7.12 * 10^{-5}$

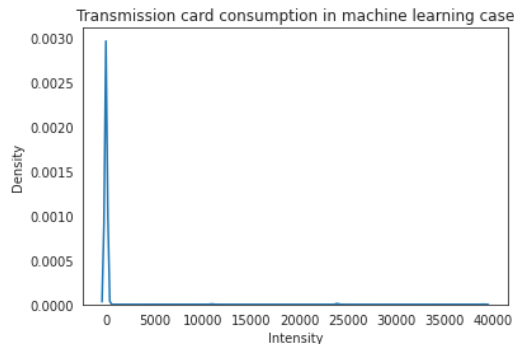
(18.28661818163327, 18.293796941822553, 18.300975702011836)



(21060.063789409185, 21337.932378854624, 21615.800968300064)

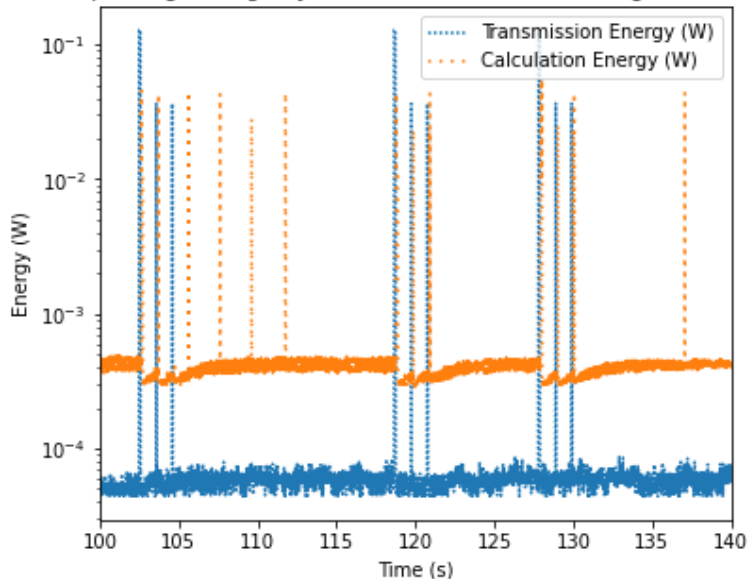


(161.61508877972426, 166.319445587194, 171.02380239466373)

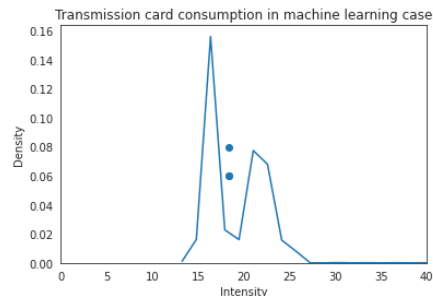


Energy consumption

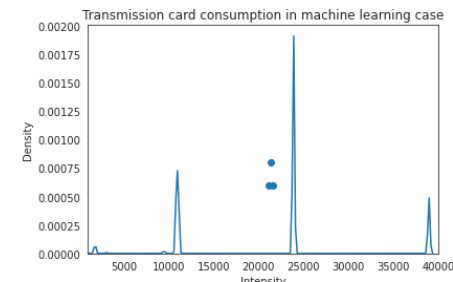
Power passing through by our electronic cards in W (against time in s)



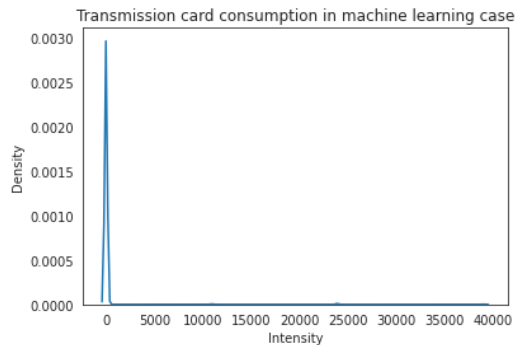
(18.28661818163327, 18.293796941822553, 18.300975702011836)



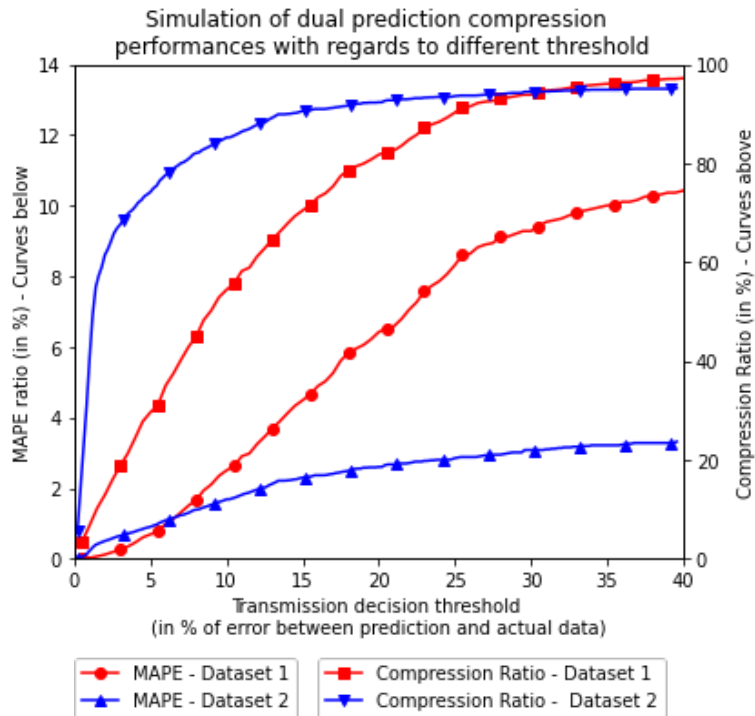
(21060.063789409185, 21337.932378854624, 21615.800968300064)



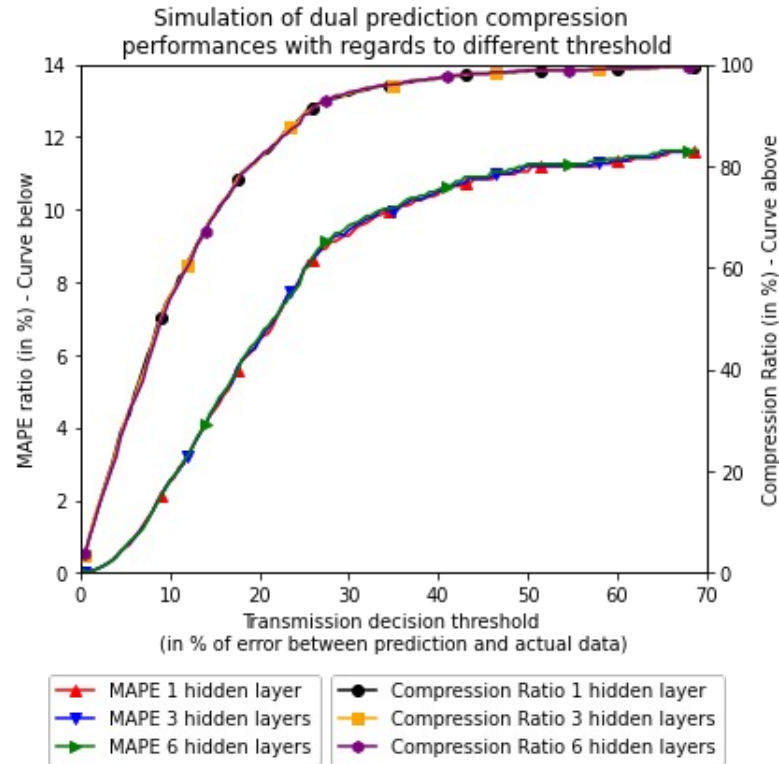
(161.61508877972426, 166.319445587194, 171.02380239466373)



Compression performance & Error rate



Compression performance & threshold



Quantification results

Comparison between Float32, Float16 and Int-8 Quantified data prediction

