SITH: Smart Internet of THings

Fabien Ferrero, Benoît Miramond, Laurent Rodriguez LEAT, Université Côte d'Azur







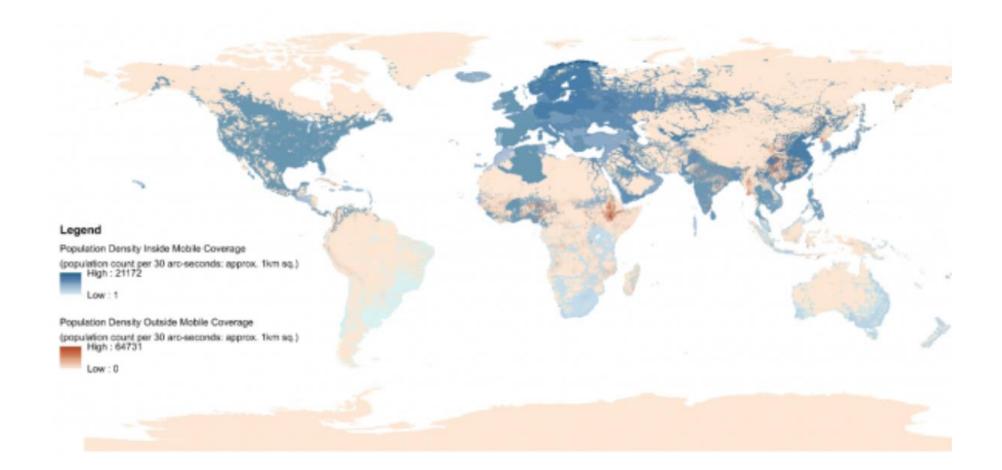
LABORATOIRE D'ELECTRONIQUE ANTENNES ET TELECOMMUNICATIONS



- Access to space for IoT
 - IoT from space: Opportunity and Challenge
 - Antenna for lot from space application
 - Prototype and Measurement
- EDGE AI
- CERN project
- Teaching
- Conclusion and perspectives

Mobile phone coverage



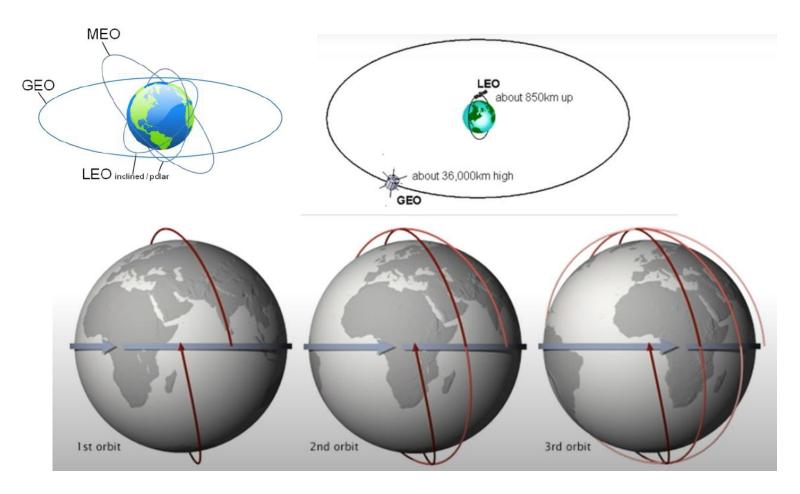


EPAR's 2014 Review of Mobile Coverage (terrestrial networks)



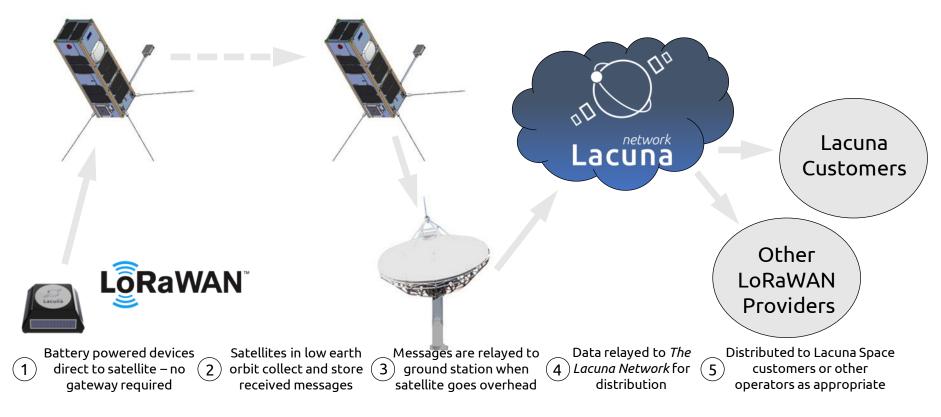
IoT from space with LEO satellite

Low Earth Orbite reduce communication distance





How does the Lacuna technology work?



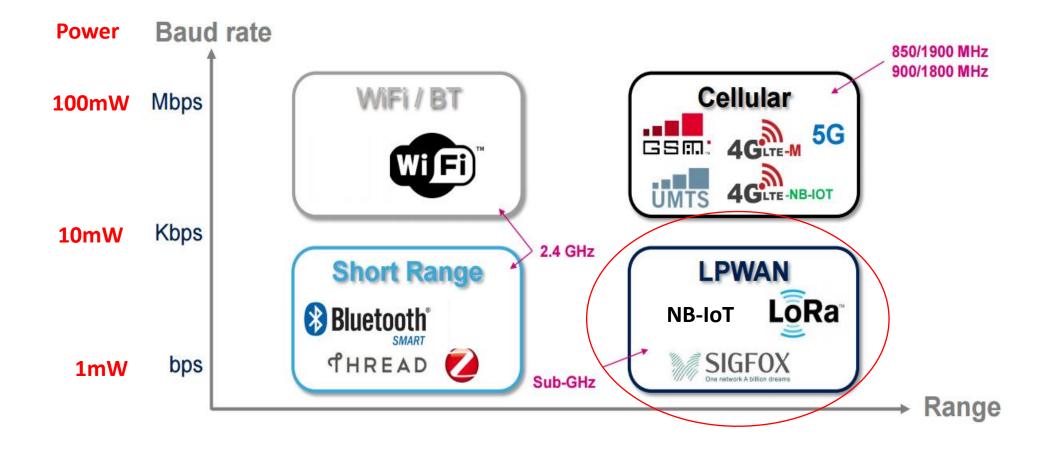
Use ISM bands : 868 or 920MHz

Lacuna



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Siths' UCA BoardRFT dev board

- The target is a RFT-AI Dev. Kit board equipped with a STM32L476RGT6 Microcontroller. This MCU is based on the ARM Cortex M4 architecture and runs at a frequency of 80 MHz. The board provides 1
 MB Flash and 128 KB SRAM.
 - LoRa SX1262 Module with CP antenna
 - Quectel L96 M33 GPS module
 - Accelerometers
 - Gyroscope
 - Magnetometer
 - 9 Axis Sensor TDK InvenSense ICM-20948 Digital,
 - PDM Microphone MEMS (Silicon) Omnidirectional SPH06
 - Air Quality Sensor Sensirion AG SGP30-2.5K
 - Optical Sensor Ambient Lite-On Inc. LTR-303ALS-01







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Sith Development Kit

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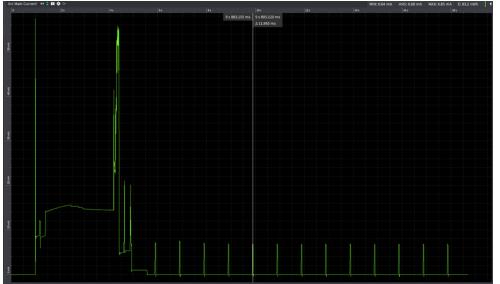
Based on LS-200 Reference Schematic :

- 868/923 RHCP antenna
- SX1262
- STM32L476 (1Mb flash)
- Ublox M8Q GPS
- Sensors : Accelerometer, Air quality, ..
- Hall sensor

Sleep power : 20uA Relay mode : 60uA

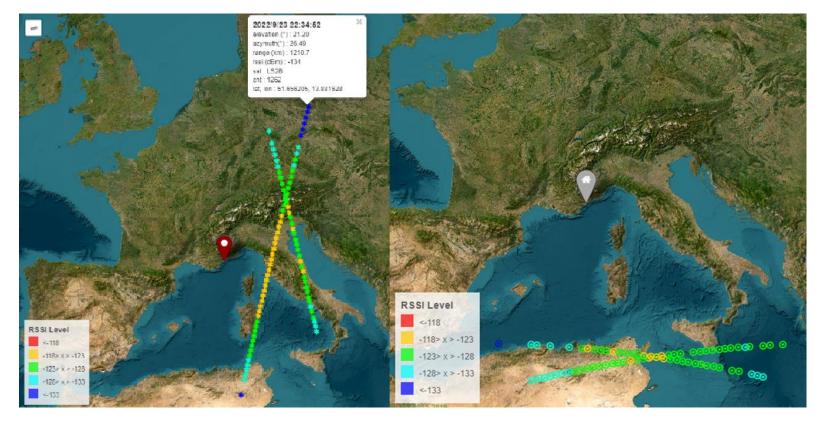
Several years of autonomy with a single battery





In the Field first results





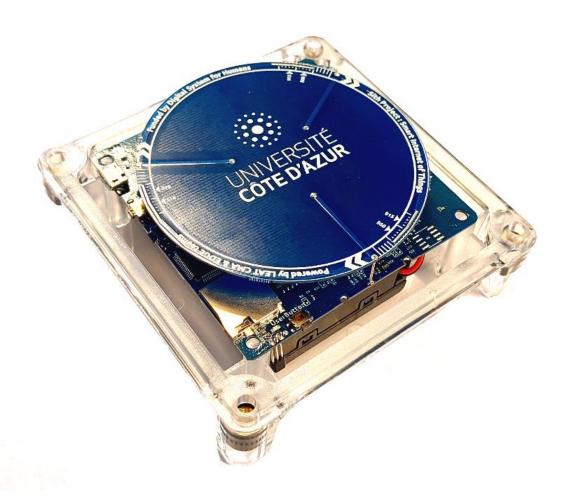
LS2D (polar orbit)

LS2C (37° inclination orbit)

Communication down to 20° of elevation

Siths' UCA BoardRFT dev board





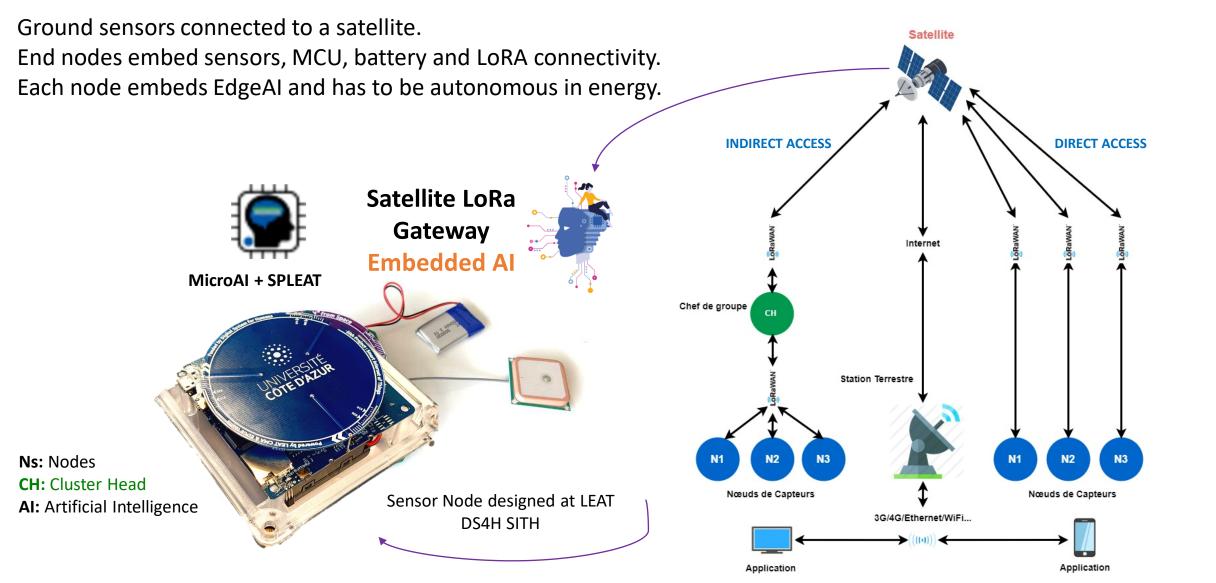
- The developed board, doted with the designed antenna becomes the first IoT object communicating with space through LPWAN
- Data throughput in the order of tens of bits per second

=> Need for AI on the edge for complexes applications



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Example of distributed AI with Satellite IoT



I. Abdoulaye, L. Rodriguez, C. Beleudy, B. Miramond, Embedded Artificial Neural Network for Data Prediction in Efficient Wireless Sensors Networks, ASPAI 2022

14



- Edge AI offers the possibility to embed near-sensor processing
- By bringing AI closer to the sensor, the goal is:
 - To reduce the amount of data to communicate
 - To lower the global energy consumption of the digital infrastructure
 - To reduce latency for decising making (close or open loop)
- It includes
 - Near-sensor classification + Data fusion
 - Distributed AI in order to enable self-organization of WSN

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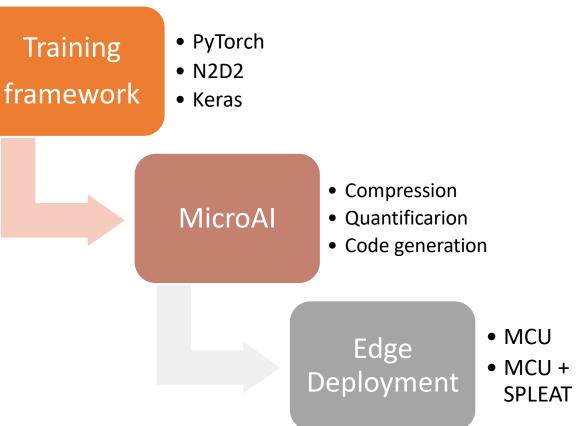


https://bitbucket.org/edge-team-leat/microai_public

Quantization and deployment of deep neural networks on microcontrollers, PE Novac, GB Hacene, A Pegatoquet, B Miramond, V Gripon, Sensors 21 (9), 2984, 2021 SPLEAT: SPiking Low-power Event-based ArchiTecture for in-orbit processing of satellite imagery, N. Abderrahmane B. Miramond, IJCNN 2022

The LEAT codesign flow for Edge AI

- Complete Solution: from Training to Edge
- Training of networks (frameworks PyTorch, Keras, N2D2)
- Embedded preparation of ANN with MicroAl
 - Quantification des SNN
 - Automatic code generation
 - Open-source: https://bitbucket.org/edge-team-leat/microai_public



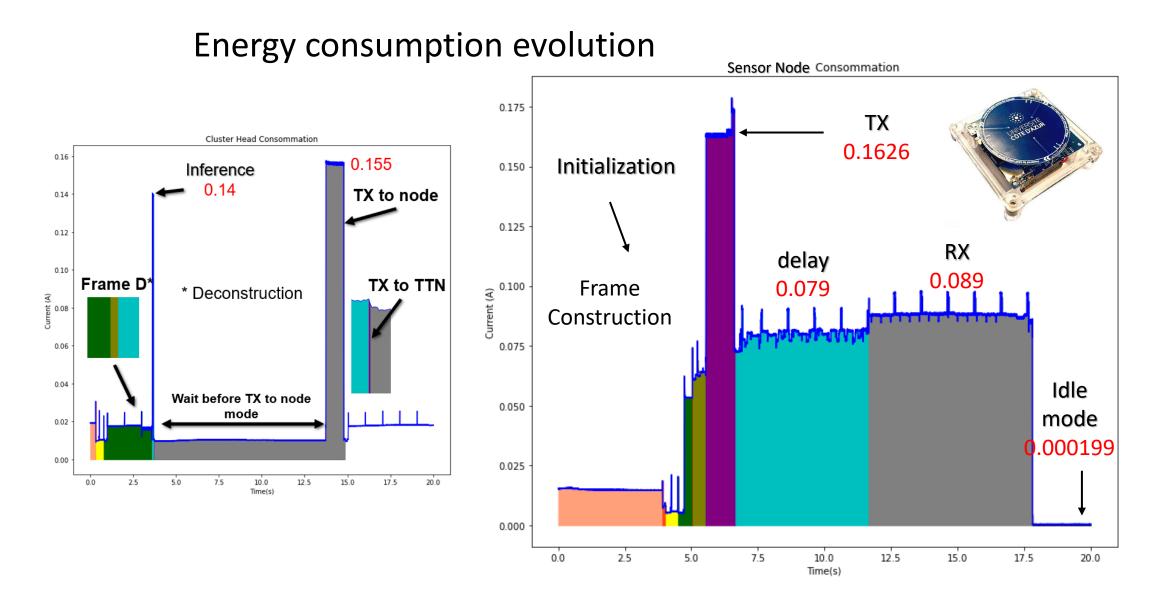




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Evaluation and Discussion of Embedded Inference







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CERN biodiversity project











SURFACE TOTALE DU DOMAINE DU CERN: 626 Ha



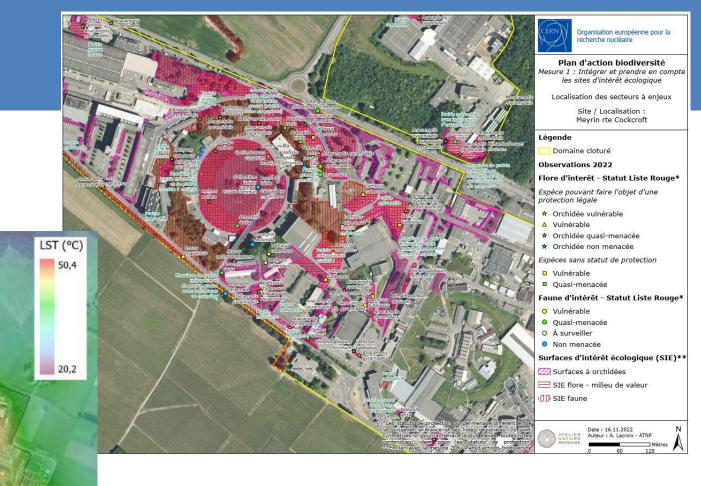
Domaine clôturé

- Meyrin (79 Ha)
- Prévessin (83 Ha)
- Point LHC
- Points SPS
- Surface totale: 211 Ha
- Surface espaces verts:101 Ha
- Surfaces voiries/parkings: 57 Ha
- Surface bâti: 39 Ha

Domaine non clôturé Bois, Forêts 136 Ha Champs 258 Ha Surface totale: 415 Ha

Présenté par Sabrina Schadegg, HSE/ENV-EP

CERN biodiversity project



CERN biodiversity project



- Objectives:
 - Deploy 200 sensors:
 - Monitoring of environmental data (air quality, temperature, etc.)
 - Detection and listing of bird species present on the site
 - Criteria:
 - Lifetime of the sensors (at least one year of autonomy)
 - Pattern recognition (environmental events, bird songs, etc.)
- Benefits:
 - To have a better knowledge of the environmental parameters to allow actions to improve the site in a concern of eco-responsibility and sustainability (reforestation, fields, work on the buildings roofs, etc.).
 - Lead future developments of the site in the respect of the environmental criteria



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Enseignements



The electronic boards funded by the SITH project enable the implementation of many concepts for various teaching modules such as embedded AI and IoT.

- Master EIT-Digital option Systèmes Autonomes (IA Embarquée)
- 4th and 5th year in computer science and electronics at Polytech Nice Sophia (Embedded AI)
- ITII 5th year (Embedded AI and communicating objects)



Etudiants du Master EIT-Digital déployant un modèle d'intelligence artificielle pour la reconnaissance d'activité humaine sur les cartes du projet SITH dans le cadre d'un cours sur l'IA embarquée

- •
- ~ 120 Students



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Conclusion



- First card to enable IoT communication to space with AI capability
- Enable collaboration between LEAT's IoT and Edge AI team
 - Synthesize our respective skills within the same platform
- Opening national and international collaboration (INPHYNI, CERN, HCMC University Vietnam, DaNang University Vietnam)
- Evolution:
 - Energy autonomy: energy harvesting (solar) + super capa
 - Increase supported standards (BLE, Wifi, S-band geostationary communication)
 - Modularity (daughter card MCU, Sensors, etc.)

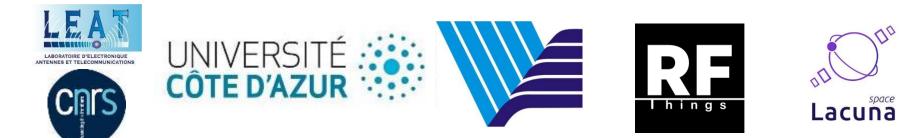


LABORATOIRE D'ELECTRONIQUE ANTENNES ET TELECOMMUNICATIONS

Laboratory of Electronics Antennas and Telecommunications



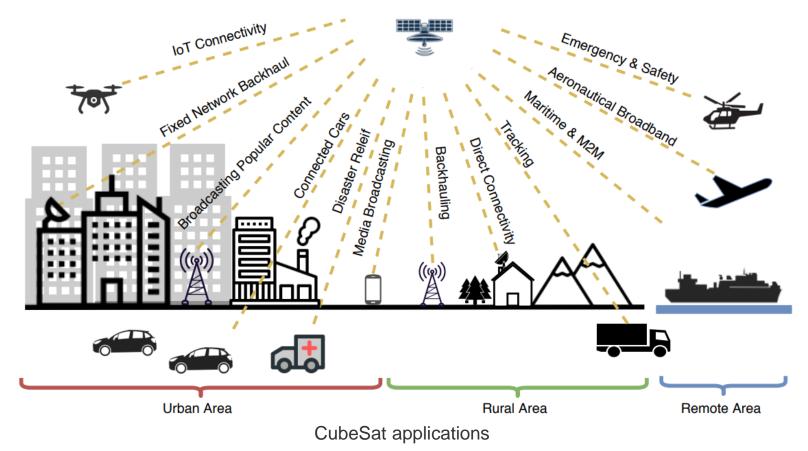
<u>fabien.ferrero@univ-cotedazur.fr</u> <u>benoit.miramond@univ-cotedazur.fr</u> <u>laurent.Rodriguez@univ-cotedazur.fr</u>



New space for IoT



CubeSat provide an "affordable" access to space



O. Kodheli *et al.*, "Satellite Communications in the New Space Era: A Survey and Future Challenges," in *IEEE Communications Surveys & Tutorials*, vol. 23, no. 1, pp. 70-109, Firstquarter 2021, doi: 10.1109/COMST.2020.3028247.

IA on the Edge : Deployment on MCU

After the network has been trained and quantized it is deployed:

- **Export the weights** of the DNN and encode them into a format suitable for on-target inference
- Generate the inference program according to the topology of the DNN
- **Compile** the inference program
- **Upload** the program with weights onto the MCU's ROM

Pierre-Emmanuel Novac, Ghouthi Boukli Hacene, Alain Pegatoquet, Benoît Miramond, Vincent Gripon. Quantization and Deployment of Deep Neural Networks on Microcontrollers. *Sensors*, MDPI, 2021

IA on the Edge : Deployment on MCU

Data acquisition and windowing float32 Training int8 Quantization-aware training with PyTorch **MicroAl General Flow** PyTorch to Keras model conversion int16 Post-training quantization C inference code generation from Keras model Deployment on microcontroller Evaluation on microcontroller

Pierre-Emmanuel Novac, Ghouthi Boukli Hacene, Alain Pegatoquet, Benoît Miramond, Vincent Gripon. Quantization and Deployment of Deep Neural Networks on Microcontrollers. *Sensors*, MDPI, 2021

Conclusion

- The combination of Edge Al and sensors
 - makes AI to the contact of the physics of the real world
 - Addresses the question of the energy consuption reduction of AI
- By bringing AI closer to the sensor, the goal is
 - To reduce the amount of data to communicate
 - To lower the global energy consumption of the digital infrastructure
 - To reduce latency for decising making (close or open loop)
- Original approach and promising results on bio-inspired AI thanks to
 - Greater sparsity
 - Event-based processing (specific neuromorphic hardware)
 - Reduced power consumption
 - And a large amount of unexplored features in the brain
- Remaining challenges for EdgeAI / Neuromorphic architectures
 - Training methods SNN are not so mature compared to standard machine learning
 - A lack of european actors in the domain of Electronic manufacturers for Al Chips
 - On-line/On-board training is still an open question related to catastrophic forgetting in ML
 - Interested ? Join 3IA Côte d'Azur ... join LEAT



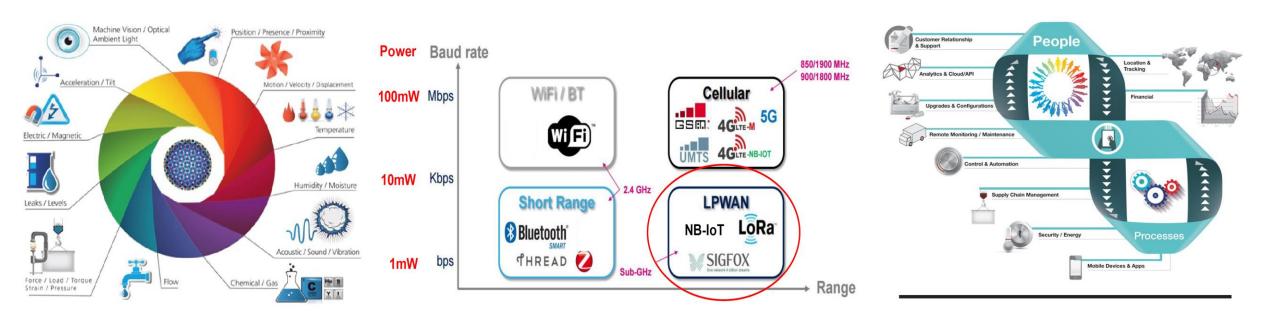
Key elements of IoT sensors



Sensors

Connectivity

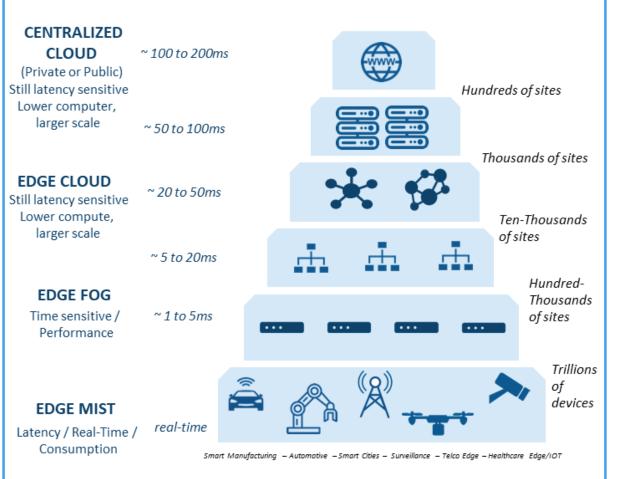
Persons & process



Captures a discrete representation of the dynamics of the physical world Transmits the sensors data through wireless communication

Provides the information to people or process the raw data into more abstract information

Edge Lines and their specific constraints

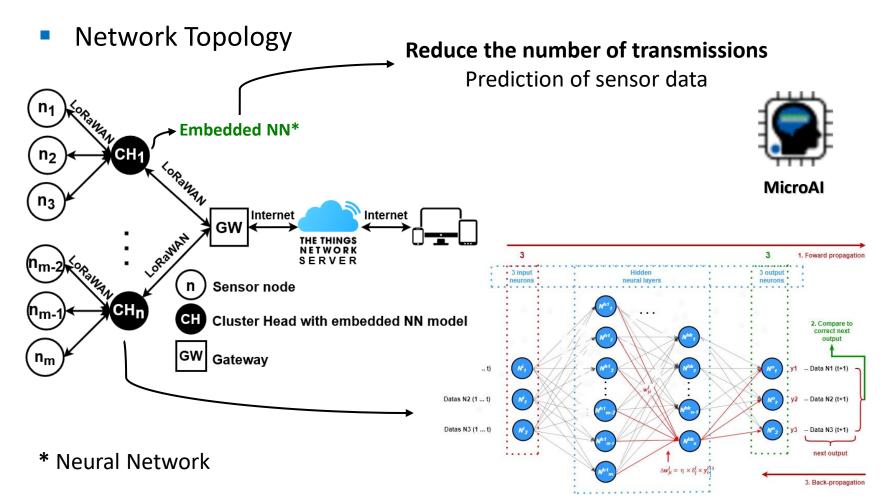


	Memory	Computation	Power	Efficiency
Edge Servers	GB	1 Tops	100 W	10 Gops/W
Gateway	MB	100 Gops	1 W	100 Gops/W
loT Nodes	Hundreds of kB	1 Gops	1 mW	1 000 Gops/W

LABORATOIRE D'ELECTRONIQUE



Overview of the global infrastructure

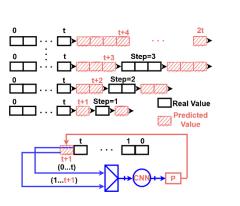


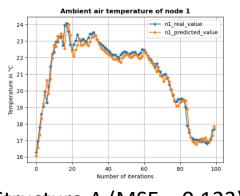
L. Lizzi et al., Solutions Antennaires Dédiées aux Véhicules Développées au LEAT

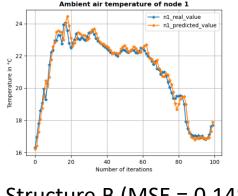
Experiments and Results

Results on different simulations

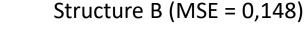
Decision on change of rate



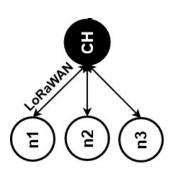


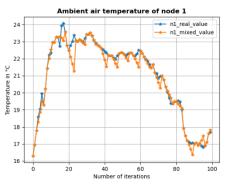


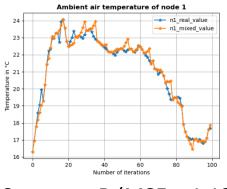
Structure A (MSE = 0,123)



Evolution of real and predicted data for sensor node 1.







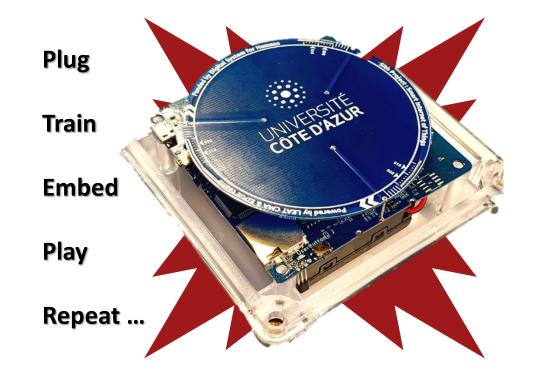
Structure A (MSE = 1,637)

Structure B (MSE = 1,136)

Evolution of real and predicted data for sensor node 1 for a step of 3 for the decision on change of rate.

EdgeAI, let's play !

The field of possibilities is only limited by your imagination



IDEX Sith project, F. Ferrero, L. Rodriguez, B. Miramond

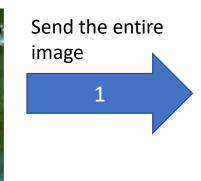














VS.



Send only the images without clouds, fire, ...



