#### IEEE International Smart Cities Conference (ISC2) September 2021

Title: Al for IoT and Smart Cities

Presenters: Kamal Singh and Guillaume Muller

**Affiliation:** University Jean Monnet, St-Etienne, France









#### Presenters

Kamal Singh



Guillaume Muller



#### Plan

- Session 1: IoT Landscape, Al
- Session 2: Federated Learning
- Session 3: Exercises
  - Installation instructions to follow in advance
  - https://github.com/GMTSE/isc2-IoT-AI-smart-tutorial/blob/main/README.md
  - If you do not have python3.6 or above and cannot install it
    - then please download the VM
    - when VM starts then please do not select factory reset

#### Session 1 Plan

- IoT Landscape
- Al for IoT and Smart Cities
  - Applications
  - Al
  - Opportunity and challenges

## IoT Landscape

#### Introduction

IoT combines different domains, the physical and digital world



 Challenge and opportunities are to create Intelligent services for our society and environment

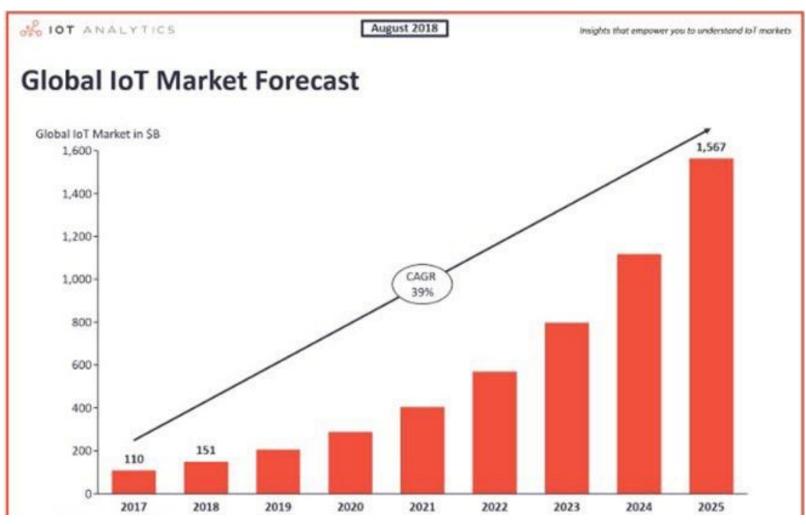
Ref: Al-Fuqaha et al., "Internet of Things: A Survey on Enabling Technologies, Protocols and Applications"

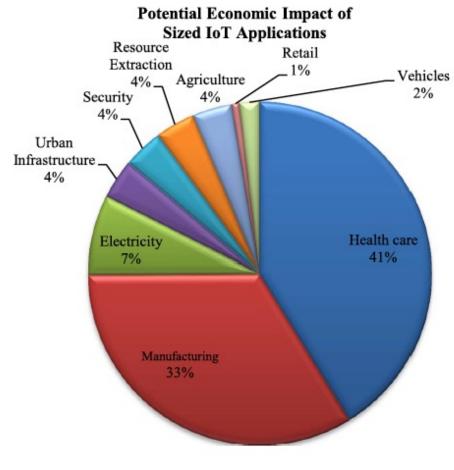
Applications



Ref: Gubbi et al., "Internet of Things (IoT): A vision, architectural elements, and future directions"

#### Market





Source: IoT Analytics Research 2018

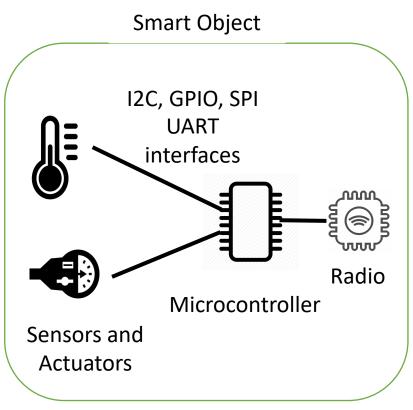
### Challenges and Opportunities

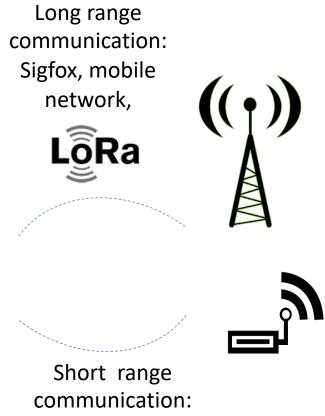
- Power consumption
- Device constraints
- Latency
- Privacy
- Security

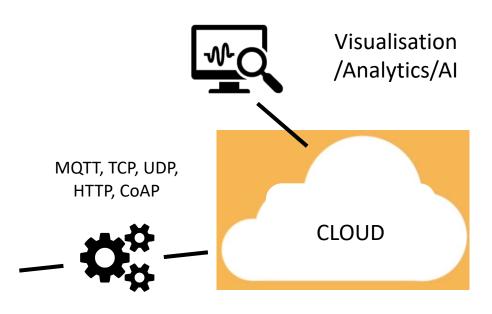




#### Chain







• • •

Zigbee, BLE, Wi-Fi,

#### IoT Landscape

- Device capabilities
- Communication protocols
- Computing

#### Plan

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#### Range of Boards and Microcontrollers

1.5GHz, 2-8 Raspberry Pi **GB RAM** 

Pycom

Arduino

• ESP8266, ESP32

Particle

Pycom: 160 MHz, 520 KB RAM, 4 MB external



MHz, 520 KB

**RAM** 



Arduino Nano 33 IoT: 48 MHz, 32 KB **RAM** 

MHz, 128 KB **RAM** 

#### Range of Boards and Microcontrollers ...

- Memory, MCU, Power consomption, Price (ESP32: 1 to 4\$!)
- Al and Image/audio based applications: Nvidia Jetson Nano, GPU with 128 CUDA cores
- Al accelarators: NPUs (Neural processing unit), ISPs (Image signal processor), and VPU (Vision processing unit, Video processing unit)



#### OS

	Contiki	TinyOS	FreeRTOS	Zephyr	Mbed	RIOT
Characteristics	Event Driven	Event Driven	Multi- threading, RTOS	Multi- threading, RTOS	Event Driven	Multi- threading, RTOS
Programming	C	nesC	C	C	C, C++	C, C++
License	BSD	BSD	Modified GPL	Apache 2.0	Apache 2.0	LGPLv2

Some other OS allow programming using microPython and circuitPython.

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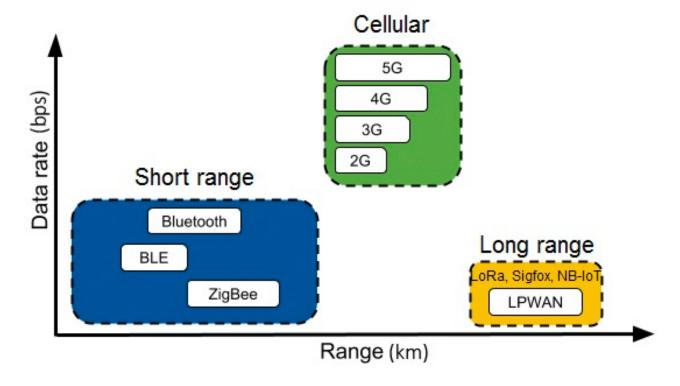
# Short Range vs Long Range Communication



Source: pngegg.com

#### Data rate vs distance

Choose a technology as a function of the application



Ref: <a href="https://www.researchgate.net/figure/Range-and-data-rate-of-different-communication-technologies\_fig1\_332187524">https://www.researchgate.net/figure/Range-and-data-rate-of-different-communication-technologies\_fig1\_332187524</a>

#### LoRAWAN: open business model

- Possible to deploy a private network as well as an operator network
- An international opensource LoRAWAN network: TTN https://www.thethingsnetwork.org/



#### LoRAWAN, Sigfox, ...: whats next?

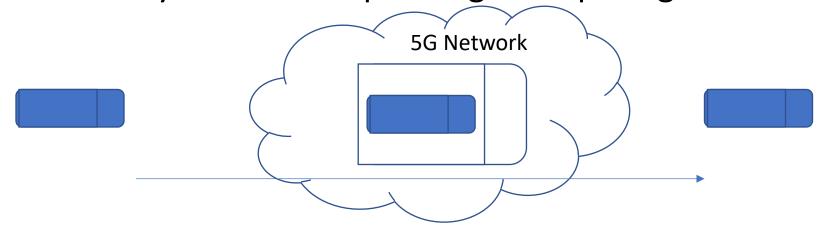
- Higher bitrates?
- IP connectivity? Header compression, RFC 9011 for LoRAWAN, a draft for sigfox, ...



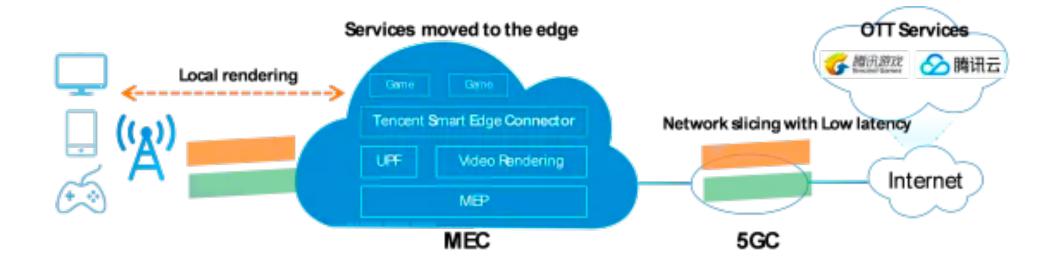
Source: www.ackl.io

#### What does 5G bring to IoT and Smart Cities?

- Private 5G networks
- Transport of other network protocols: Ethernet, CAN
- eURLLC for TSN (Time Sensitive Networking)
- Flexibility with the help of virtualisation and disaggregation
- Low latency with the help of Edge computing



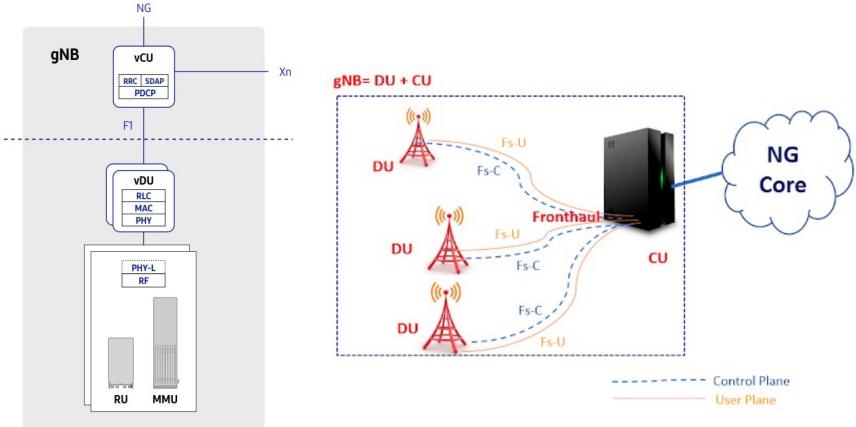
#### Multi-Access Edge Computing (MEC)



#### Disaggregation in 5G

- gNB can be cut
- Further CU can be cut into CU-CP and CU-UP

• ..

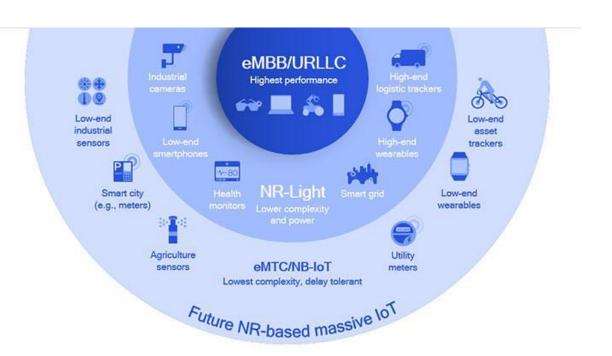


Ref: techplayon.com

Ref: <a href="https://www.5g-networks.net/5g-technology/virtualised-and-disaggregated-5g-nr-vran-architecture/">https://www.5g-networks.net/5g-technology/virtualised-and-disaggregated-5g-nr-vran-architecture/</a>

#### What is next for IoT and 5G/Beyond5G?

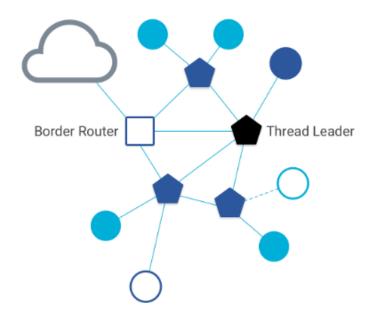
- New Radio Light (NR Light) for energy efficiency
- Massive IoT
- Support for AI applications
- eURLLC
- Precise localisation



Ref: <a href="https://www.qualcomm.com/news/onq/2020/05/12/5g-here-whats-next-internet-things">https://www.qualcomm.com/news/onq/2020/05/12/5g-here-whats-next-internet-things</a>

#### Short Range Communications

- Zigbee, Bluetooth Low Energy (BLE), Zwave, ANT, NFC ...
- Relatively new protocol: Thread

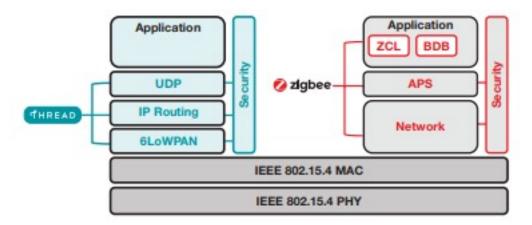


#### Thread





- Launched in 2014 and open source in 2016. Thread group alliance includes Google, Samsung, Nest, Freescale et ARM, NXP, etc.
- Native support for IP. In comparison, Zigbee should translate its address to IP using an intelligent gateway
- Ease of authentication based on smartphone

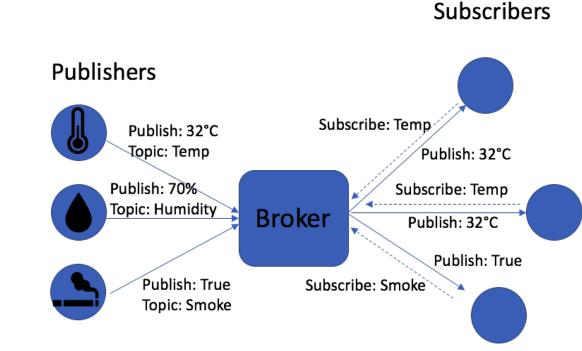


Ref: https://e2e.ti.com/blogs /b/process/posts/thread-vs-zigbee-what-s-the-difference

# Publish-Subscribe and REpresentational State Transfer(REST)

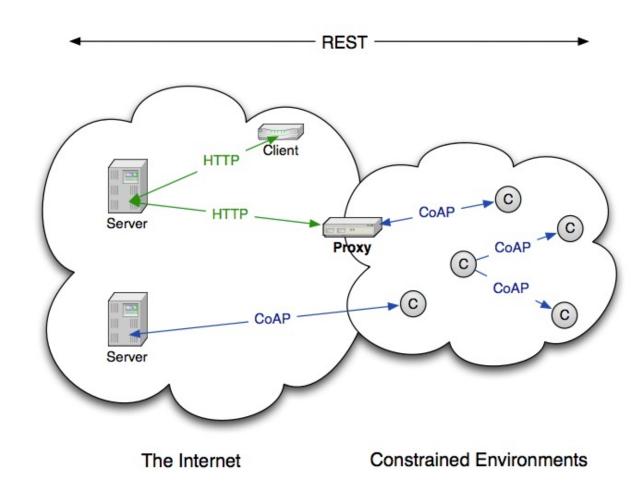
## MQTT (Message Queuing Telemetry Transport)

- Publish to Topics: myhome/firstfloor/room1/tempsensor
- Subscribe using wildcards (+ or #)
  - All sensors: myhome/#
  - All tempsensors on firstfloor: /myhome/firstfloor/+/tempsensor
- Advantages/disadvantages
  - Sender and receivers are decoupled
  - Needs TCP which can be heavy for some sensors
  - A variant MQTT-SN uses UDP



#### CoAP (Constrained Application Protocol)

- REST APIs are very popular
- 4 bytes header, UDP, SMS (TCP)
- Security with DTLS
- Discovery
- Option Observe: useful for asynchronous observations



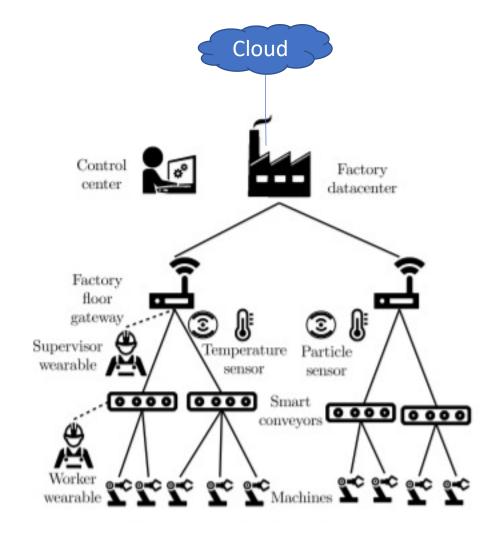
Source: ARM CoAP tutorial

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#### Edge Computing

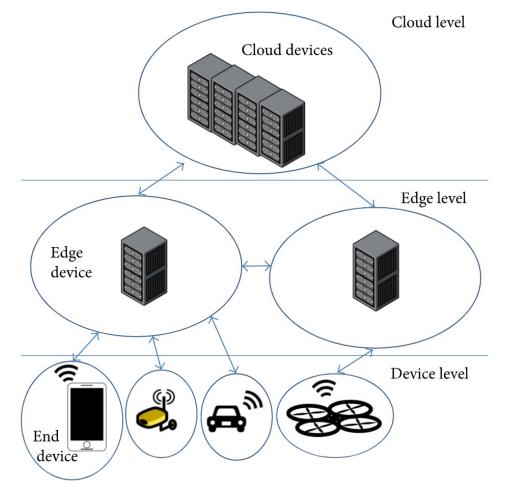
- Need for lower latency
- Local intelligence is needed
  - Instead of messages travelling to and fro from cloud



Ref: EDR: A Generic Approach for the Distribution of Rule-Based Reasoning in a Cloud-Fog continuum

#### Edge Computing Research challenges

- Heterogeneous environments
- Resource management
- Scalability
- Reliability, Fault tolerance
- Security, Trust



Ref: Toczé et al. "A Taxonomy for Management and Optimization of Multiple Resources in Edge Computing"

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#### Al based Applications

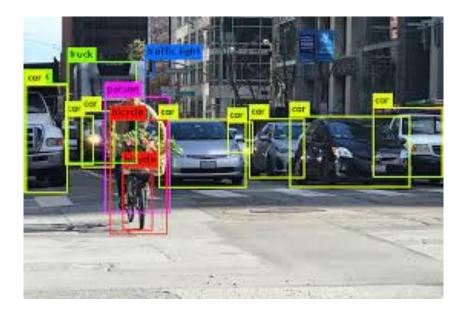
#### Camera

- Crowd density estimation, cooling depending on no. of persons or weather, social distancing
- Urban emergency, traffic management, driver attention
- Face mask detection
- Activity detection

#### Audio

- Detecting running equipment
- Anomaly detection (car engine, machine going bad, elevator)
- Accelerometer
  - Anomaly detection, activity detection





Source: YOLO software

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### Al



#### Symbolic

- High level symbolic representation, semantics, human readable, logic
- Expert systems, reasoners, etc.
- Statistical
  - Statistical and mathematical tools, data mining
  - Machine learning

Recently statistical approaches have shown good performances, i.e., deep learning

Nevertheless both are needed

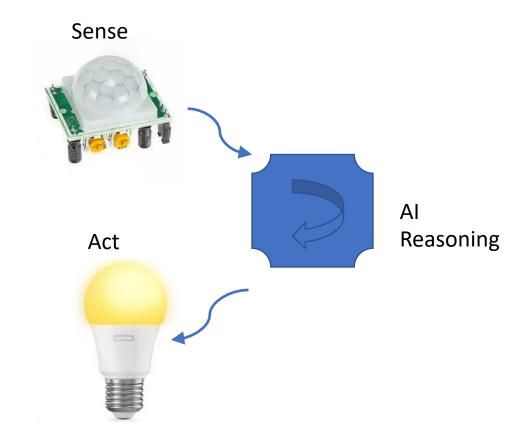
 Shifting away from symbolic AI may be a shift from explainable AI – Noam Chomski

### Plan

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  - Al
    - Symbolic
  - Opportunity and challenges

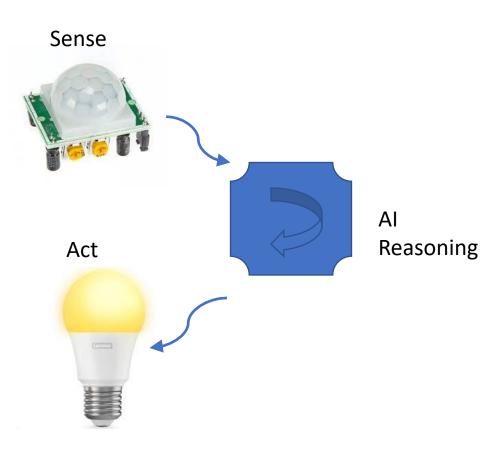
# Symbolic AI: Rule based reasoning

• Example: If luminosity < 50 and presence == 1 then light the lamp



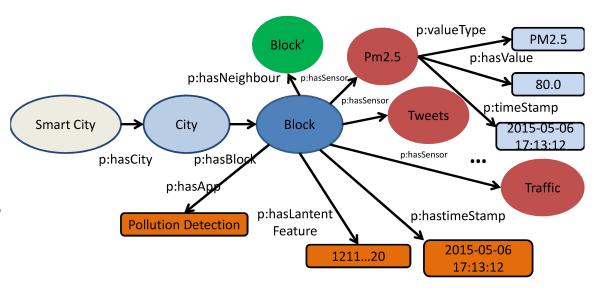
# Problems to make it automatic

- How will AI recognise required concepts?
  - luminance, Presence, etc.
- Semantic interoperability between different devices?
  - pir or presence sensor?
  - Will my definitions work with other AI?
- Context?
  - Is the presence detected in the same room?



## Knowledge representation, Semantics

- Ontologies: specify the concepts and relations of concepts within a domain
  - machine understandable and human readable specifications
  - being largely adopted in the industry for interoperability and federation
- Knowledge graphs: using ontology as framework, we can build a graph of real data
- Reasoning and Querying can be done on graph data



Ref: Zhang et al. Semantic Framework of Internet of Things

for Smart Cities: Case Studies

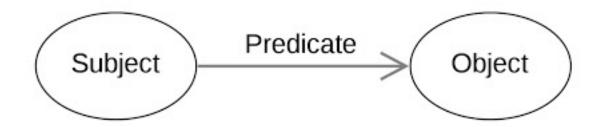
## Some technologies and concepts

REST allows loose coupling between devices and applications

• URI: Uniform resource identifier to reference a resource coap://192.168.1.52/capteur1/temperature?max value&date=20131206

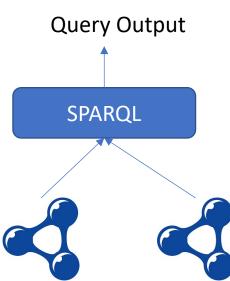
JSON, JSON-LD data formats{ "temperature" : 25} a JSON object

## Resource Description Framework (RDF)



- W3C standard data model
- Triple: subject, predicate (property), object
  Paris is capital of France (p:Paris p:Capital p:France .)
  p: is prefix abbreviation for "http://example.org/data.ttl#"
- Multiple triples can be combined to form a graph

# SPARQL (SPARQL Protocol And RDF Query Language)

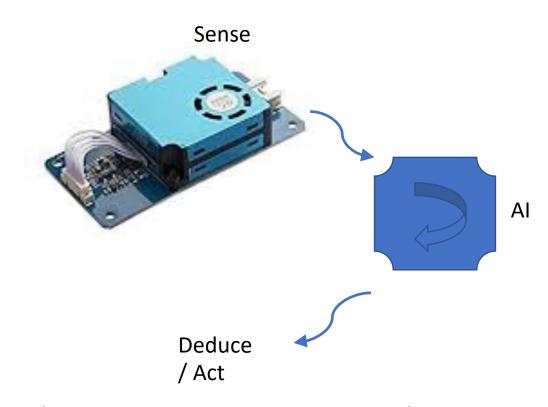


You can test different queries on a big pubic knowledge graph: <a href="https://dbpedia.org/sparql">https://dbpedia.org/sparql</a>

# Rule based reasoning for inference and deductions

If PM2.5 is more than 50 then city is polluted

IF PM2.5 is more than 150 then city is very polluted



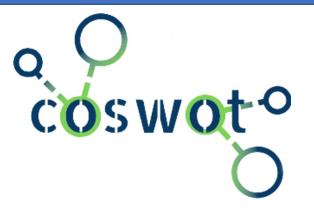
Reasoners: RETE, RDFox, Hermit, Jena (contains many modules)

# Constrained semantic web of things (CoSWoT)

#### French national project to design WoT application platform able to:

- (1) use graph-based knowledge models
- (2) distribute and process **reasoning** tasks among heterogeneous nodes, including **constrained devices**

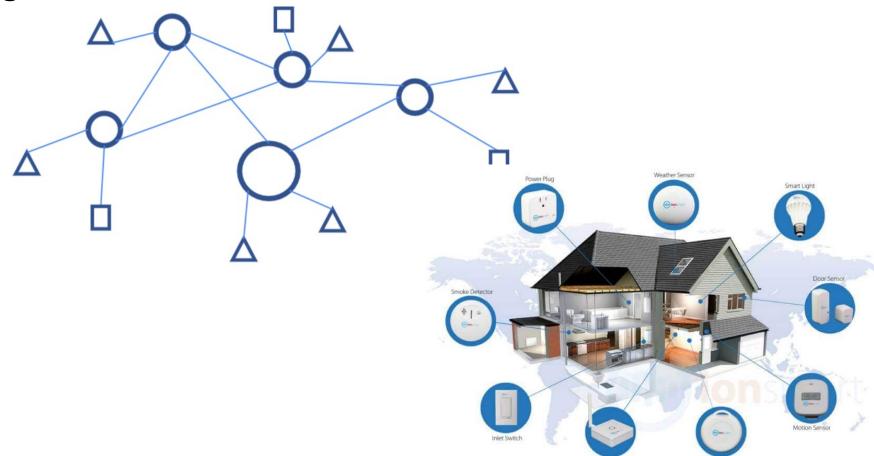
Will enable the development and execution of traceable and decentralised smart WoT applications despite the heterogeneity of devices



### Use cases

• Smart buildings

Agriculture

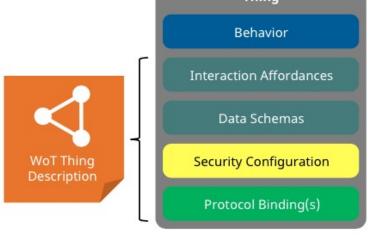


### CoSWoT

- We are Hiring! Postdoc on distributed reasoning for web of things
  - Requirements: PhD in computer science
- Email: kamal.singh@univ-st-etienne.fr
- For info on the project:
- https://coswot.gitlab.io/

# Frameworks and ontologies

# W3C Web of things



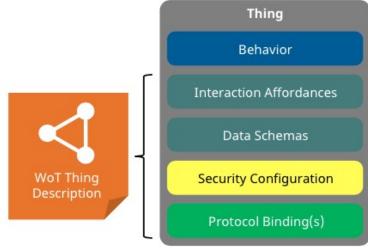
Thing Thing Consumer ¢..... Exposed Thing

- WoT vs IoT: WoT works on application layer while IoT on Network Layer.
- The Web of Things (WoT) is application layer software that allows real-world objects to be a part of the Web.
- WoT architecture standardised by W3C: thing, Things Description, consumer, ...

# Web of Things: Things Description (TD)

- A formal description of meta-data and interfaces of a Thing.
- Defines ways of interactions with a thing:
  - properties
  - actions
  - events
- Example a thing (1 RGB LED, a temperature sensor)
  - properties: temperature, RGB LED color
  - action: change color ...





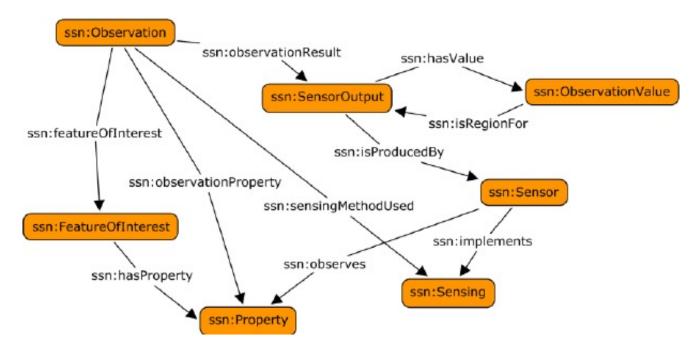
# Web of Things: Things Description (TD)



```
"@context": "https://www.w3.org/2019/wot/td/v1",
"id": "urn:dev:ops:pysense-rgb-led-1234",
"title": "PYSENSE-RGB-LED", "description": "RGB LED on Pysense",
"securityDefinitions": {"nosec sc": {"scheme": "nosec"}},
                                                                             HTTP GFT or PUT to
"security": ["nosec sc"],
                                                                             http://192.168.0.17/pro
"base": "http://192.168.0.17/"
"properties": {
                                                                             perties/color
   'color": {
     "title": "Color", "description": "The color of the LED",
     "links": [{"rel": "property", "href": "/properties/color"}],
     "@type": "ColorProperty", "type": "integer"
    "temperature": {
     "title": "Temperature", "description": "The temperature sensor value",
     "links": [{"rel": "property", "href": "/properties/temperature"}],
     "@type": "Temperature", "type": "number"
"actions": {
  "changecolor": {
     "title": "Change Color", "description": "Change the color of LED",
     "links": [{"rel": "action", "href": "/actions/changecolor"}],
     "input": {"required": ["color"],
     "properties": {"color": {"minimum": 0, "maximum": 16777215, "type": "integer"}},
     "@type": ["ColorControl"], "type": "object"
```

## Other frameworks and ontologies

- oneM2M (machine to machine)
- SAREF (devices, sensors, actuators)
- OPC Unified Automation (architecture for industrial automation)
- SSN/SOSA (sensors, observations)



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## Machine Learning

- Supervised, unsupervised, semi-supervised
- Traditional: SVM, RF, Xboost, Reinforcement learning, etc.
- Neural Networks
  - Deep Learning,
  - Auto-encoders, VAE
  - Deep Reinforcement learning
- Distributed learning, Federated learning

# ML details in session 2

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# Research challenges for Knowledge Representation and Reasoning

- Embedded reasoning
  - RDF, Reasoners, SPARQL can be heavy for embedded devices
  - For the moment they are realised on the cloud
- Edge computing platforms are needed
- Can they be implemented on embedded devices?
  - Project CoSWoT
- Can we do efficient reasoning on streams and real time data?
  - Project CoSWoT

### Embedded ML Challenges

- How to fit ML models into embedded devices?
  - Quantisation / sometimes floating point is not supported
  - For example see Microsoft Research: <a href="https://github.com/microsoft/EdgeML">https://github.com/microsoft/EdgeML</a>
- Problem of obtaining labels
  - Semi-supervised or unsupervised techniques
  - VAE, anomaly detection
- Privacy
  - Federated learning

# Thanks!