IOT Online Course

Fundamentals of IoT

F-IOT-2b: Understanding IoT Devices, Architecture & Ecosystem

Prof. Congduc Pham http://www.univ-pau.fr/~cpham Université de Pau, France





T2I team







• http://diy.waziup.io			IOT COURSES	
Introduction Arduino LoRa IoT x + ← → C		WAZIUP IoT Courses		
		 For users who wants to gain knowledge on IoT in a step-by-step lecture mode, we have defined the following curriculum with materials from both existing sources and specific materials produced by WAZIUP/WAZIHUB project. «Fundamental of IoT» 		
Introduction V	INTRODUCTION	F-IOT-1a: What is loT ?	F-IOT-1b: Introduction to Basic Elec	etronics
 Arduino IDE Sensors Advanced boards Advanced boards MQTT Node-RED LoRa communication WaziDev board Solution Lab 	This online tutorial on Arduino, Sensors, IoT and LoRa tech by University of Pau, France, in the context of the WAZIUP funded by the European Union in the H2020 research prog this online tutorial is to provide comprehensive and guided in training, hackathons, bootcamps, entrepreneur's days WAZIUP/WAZIHUB across Africa. The main contributors a Muhammad Ehsan and Congduc Pham ⁶² . Our main current networks and IoT but this tutorial first start with basic of A programming to understand sensing systems that are the f Internet-of-Things (IoT) concepts. Then in a second step, w protocols and technologies with a focus on LoRa radio tech low-cost. Ione-rance and energy-efficient IoT devices.	 HAZI Quick introduction to HAZI I IoT and Big Data Platf Intel IoT What Does The Inte Edureka Internet of Things (I Geospatial IoT IoT- What is In IBM Think Academy How It V F-IOT-1b: Introduction to I HAZI I Introduction To Basic Introduction To Basic Electroni 	 HAZI Introduction To Basic Electronics - Introduction To Basic Electronics - Makers Basic Electronics - Instructables HAZI Introducing physical sensors, part HAZI Introducing physical sensors, part F-IOT-2: IoT ecosystem and hardwa HAZI INF-IOT-2a: Wireless Communication	- Spaces 1 2 re m Essentials
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✓ Collapse sidebar		Bacheling Preduktyoney Crowth Master	• WAZI - F-IOT-4: WAZIUP Open Technolog	gies for Low-cost IoT







...shows communicating objects





IoT=interactions with physical work care work and the second seco





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Radio-Frequency Identification (RFID)
Near Field Contact (NFC)















EXEZ

• Radio-Frequency Identification (RFID)

• Near Field Contact (NFC)





Interaction: always complex?





9







Pictures from WiThing, https://www.withings.com/eu/fr/products/body











... but IoT usually means cloud data arithmeans





IoT added-values come from interactions and linked data!





Prof. Congduc Pham http://www.univ-pau.fr/~cphar («WAZIUP»)

(WAZHUD)







Pictures from ArchitectCorner

IoT cloud and visualization tools (WARZIUP)



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🕺 ιοτ 🗛	NALYTICS	Insights that empower you to u	Inderstand IoT market		
	Top 10 IoT Application areas 2020				
		Global share of Enterprise IoT projects ¹	<u>Trend²</u>		
<u>1 4</u> 2	Manufacturing / Industrial	22	2%		
2	Transportation / Mobility	15%	\bigcirc		
3	Energy	14%	\bigcirc		
4) 🗮	Retail	12%	\bigcirc		
	Cities	12%	٢		
6 🐶	Healthcare	9%	\bigcirc		
7	Supply Chain	7%	\bigcirc		
8 🛒	Agriculture	4%	\bigcirc		
9 📖	Buildings	3%	٢		
10 📩	Other ³	3% N = 1,414 proj	ects		
Note: 1. Based on 1,414 share of all projects has	publically known IoT projects (not including consumer lo declined, not the overall number of projects, 3, Other inc	I projects eg smart home, wearables, etc.) 2. Trend based on relative comparison with % of projects in the 2018 IoT Analytics IoT project list e.g., a downward arrow ludes IoT project from Enterprise & Finance sectors. Source: IoT Analytics Research - July 2020	w means the relative		



- Infrastructure monitoring, Security & Safety
- Continuous process improvement, Process automation, Process optimization
- Smart logistics management, remote management, tracking,
- Connectivity to back-end system, integration of smart tools, Interoperability
- Data analysis, Supply Chain Optimization, Predictive maintenance



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Industrial Internet of Things











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Irrigation



Livestock farming



Fish farming & aquaculture



Logistic, Storage, Asset Tracking





Fresh water



Is IoT the solution for your problem?

«WAZihub»

Q: How get real-time position of all city buses?





Is IoT the solution for your problem?



Q: How get real-time position of all city buses?









Is IoT the solution for your (WARZiup) problem?

Q: How to enable municipal street sweepers to report illegal dumping, leaking pipes and emergencies?



I know! I know !

A: Give them a smartphone and they can use it for reporting!

24



Is IoT the solution for your (WARZiup) problem?

Q: How to enable municipal street sweepers to report illegal dumping, leaking pipes and emergencies?







ITU Telecom World 2018 Phathwa Senene at MTN booth



IOT TECHNOLOGY ? CONCEPT ?





• IoT device can be viewed as a simple Embedded System



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Low-cost microcontroller boards

«WAZiup» «WAZihub»





Micro-processor Micro-controller

29





Moisture/ Temperature of storage areas









Technology	2G	3G	LAN
Range (I=Indoor, O=Outdoor)	N/A	N/A	O: 300m I: 30m
Tx current consumption	200-500mA	500-1000mA	100-300mA
Standby current	2.3mA	3.5mA	NC





Low-power & long-range radios (WARZiup) for IoT systems: LPWAN networks (WARZihub)



Energy consumption comparaison HORIZ N 2020



connection if not in used







The price of pay for LPWANs!



- LoRa has very low throughput 200bps-37500bps (0.2-37.5kbps)
- WiFi 802.11n: 450 000 000 bps (450Mbps)
- WiFi 802.11g: 54 000 000 bps (54Mbps)
- Bluetooth3&4: 25 000 000 bps (25Mbps)
- Bluetooth BLE: 2 000 000 bps (2Mbps)
- <u>3G/4G</u> : 20Mbps-200Mbps
- LoRa 200bps-37500bps (0.0002-0.0375Mbps)
- 3G/LoRa ratio: 20,000,000bps/200bps=100000!



General LPWAN IoT architecture



- With increased range, LPWAN are mostly gateway-centric
 - IoT gateways are connected to Internet
 - They forward data from IoT device to Internet Servers





Low-cost, general-purpose hardwarezhub»



WHAT IS ARDUINO?

Arduino is an open-source electronics platform based on easy-to-use hardware and software. It's intended for anyone making interactive projects.



ARDUINO BOARD

Arduino senses the environment by receiving inputs from many sensors, and affects its surroundings by controlling lights, motors, and other actuators.

-	_	_	

void }	<pre>setup() {</pre>
void }	loop() {

ARDUINO SOFTWARE

You can tell your Arduino what to do by writing code in the Arduino programming language and using the Arduino development environment.











DUINO PRO MINI



Microprocessors & Microcontrollers (Waziup)

- A microprocessor unit (MPU) is a processor on one silicon chip
- A microcontroller unit (MCU) is a microprocessor with some added circuitry on one silicon chip
- Microcontrollers are used in embedded computing and most IoT devices are based on microcontrollers



From "An Embedded System Overview" by Dr. Eng. Amr T. Abdel-Hamid


From µcontroller to µcontroller board hub»

• A μ controller can be standalone...





- (Single chip)
- But, it is usually mounted on a board with additional electronics parts
 - Leds, Voltage regulators
 - \odot Easy access to pins
 - Reset button
 - Serial-USB interface

 Image: constrained of the state of the

Ground

3.3V regulated

voltage

Programming Header

Raw input voltage

Understanding imple analog sens (Laziup)

 Analog sensors provides a voltage output that varies according to a physical parameter, e.g. temperature, humidity, luminosity,...







Vcc is typically 3.3V. Microcontrollers have Analog/Digital (A/D) converter to map a voltage to a numerical value. A/D with 10-bit resolution give values in [0, 2¹⁰-1] = [0, 1023]

If 0=0V and 1023=3300mV then 3300mV/1024=3.22mV is the granularity of the measure

A digital value of 100 means 100*3.22mV=322mV If the sensor output is 10mV/1°C then the physical temperature is 322mV/10mV=32.2°C



(«WAZiup») («WAZihub»)







GND

// sensor output connected to A0 analog pin value = analogRead(A0); // now need to convert to Celcius degree And converting into Celcius Temp = value * 3300.0/1024.0; // 3300/1024=3.22mV Temp = Temp / 10; // 10mV means 1°C // now process and transmit the data





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- Cheap, open, and easy to use/program
- Huge developer communities
- Hardware is not the main important issue
- Hardware is nothing without software libraries!



Generic IoT v.s. highly specialized

- Build low-cost, low-power, generic IoT platform
- Methodology for low-cost platform design
- Technology transfers to user communities, economic actors, stakeholders,...



A simple temperature sensor example hub»





Wakes-up every 10min, take a measure (temp) and send to gateway



5µA in deep sleep mode, about 40mA when active and sending

More than 1 year with 1 measure/10min

Can run several years with 1 measure/1h

WaziDev kit: IoT in-a-box!





Included

- LiPo battery regulator accepting solar panel input
- Battery level monitor
- 2 high-current control pin
- GND rail

WaziDev

- ATMega328P, 8MHz, 3.3v
- FTDI chip
- RFM95W LoRa module
- Integrated antenna











PRIMA Low-cost soil moisture device





A soil temperature sensor can be added





• "Intelligent Irrigation in-the-box", "plug-&-sense"

• From idea to reality!











IOT

Unleash the power of loT data !

2021, billions of lot devices are deployed worldwide

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IoT added-values come from interactions and linked data!





Prof. Congduc Pham

(WAZHUD)



Integrating multiple data sources (WARZiUP)

Solo Staking APR

4.27%

4.71%

Next RPL Rewards Checkpoi

RPL Collatera

72.83%

Validator Activity

Attestation Delay

V Available Updates Swap Space Usage 22 Attestations 01:30:04 No Ω lised RAM Proposals 894 MIB 3.42% 41.4 °c 9.96 GIB 62.8 GIB RAM Usag CPU Usa 15.9_s 72 I/O Wait Tim Disk 2 Space Used Your Beacon Chain Earnings No data 64.01989 FTH Total Net I/O 0.0110 ETH Vour Validator APR Your Validator Share 44.1 GiB 32.0110 ETH 59.8 GiB Next RPL Rewards Checkpoin 08/18/2021, 09:26:36 PM Total Effective RPL Staked Approx. Rewards from Next Checkpoint Fffective RPL Stakes 14.076.15 187,144.55 175,944.55 114.87 RPL 3992.01 RPL APR for Next RPL Checkpoint rETH Price (ETH / rETH) RPL Price (ETH / RPL) 350.08% Total RPL Rewards Earned 1.003688 0.005838 400.31 RPL Node Wallet Balance 275 10.03% 401



Searching for IoT data

- Searching for information is a tough issue
 Web search engine: Google,...
- If you seek for an information, for instance the soil humidity condition in a particular farm, then you need to know where to look
- When there can be billions of IoT nodes providing large variety of data, it is difficult to find your way!
- Although sensors' data can eventually be accessed with traditional methods (web services, HTTP/REST API, ...) IoT calls for a more "automatized" and "simplistic" approach

(«WAZiup») («WAZihub»)









From "search for info" to "get the info"









Use broker nodes to manage topics
 UPPA/Duboue/S25/temp, UPPA/Duboue/S25/hum













• http://www.hivemq.com/demos/websocket-client/

📧 HIVE MQ			
Connection			
Publish			~
Topic booster_pau/test	QoS 0 v	Retain	Publish
Message			
hello again			
			le le

B HIVE MQ				Websockets Client Showcas		
Connection					connected	\approx
Publish		005	Potain	*	Subscriptions	~
testtopic/1		0 -		Publish	Add New Topic Subs	cription
				i.	booster_pau/test	~
_						
Vessages				~		
2021-09-13 21:56:57 hello again	Topic: booster_pau/test		Qos: 0			
2021-09-13 21:56:04 how are you	Topic: booster_pau/test		Qos: 0			
2021-09-13 21:55:47 hello	Topic: booster_pau/test		Qos: 0			

MQTT implementing social media (WAZiUP)

- It is very easy to implement a social media app using MQTT
- WhatsApp-like example
 - Define MQTT topic per phone number
 - Alice: myWhatsApp/0655667788
 - Bob: myWhatsApp/0611223344
 - To receive/send message
 - Alice subscribes/publishes to myWhatsApp/0611223344
 - Bob subscribes/publishes to myWhatsApp/0655667788
 - To create a group
 - Alice creates a group waziup-iot
 - •myWhatsApp/0655667788/waziup-iot
 - To join(publish) on(to) the group
 - Subscribe(publish) to myWhatsApp/0655667788/waziup-iot



0655667788

0611223344

Alice



Developing without programming



- End-users are not necessarily computer science experts nor high-skilled programmers
- Use graphical tools to build data processing flows, allowing intuivive connection from data producers to data consumers







- Node-RED is a programming tool for wiring together hardware devices, APIs and online services, e.g. clouds of various types
- provides a browser-based flow editor to wire together flows with a wide range of nodes





Node-red enabled IoT gateway



 Messages received on the IoT gateway can be injected into a Node-Red flow, allowing complex data processing to be defined by end-users]ThingSpeak





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IOT BACKOFFICE



Sources: McKinsey Global Institute, Twitter, Cisco, Gartner, EMC, SAS, IBM, MEPTEC, QAS

69

IBM



But also how to analyse the data (WARZI)

- What is the meaning of the collected data?
- Example with farming
 - What is interesting for farmers?
 - \odot Fertility detection
 - Eating/Ruminating time for welfare
 - What data can be easily obtained?
 - \odot accelerometer data with neck-mounted collar
 - How to detect relevant event from these data?







• Traditional statistic methods still valid, and useful!







• Traditional statistic methods still valid, and useful!



Use the full power of the Internet!



- IoT data are pushed on Internet data clouds
- Computing resources using Virtual Machines are obtained from Internet Computing clouds
- Parallel processing
- Optimized libraries
- Web tools to orchestrate




The Big Data Landscape!

(«WAZİUP») («WAZihub»)



		DATA SOURCES & AP					DATA RESOLIDCES	
		DAIA SOURCES & AI	13				DATA RESOURCES	
DATA MARKETPLACES	FINANCIAL & ECONOMIC DATA	AIR / SPACE / SEA	PEOPLE / ENTITIES	LOCATION INTELLIGENCE	OTHER	DATA SERVICES	INCUBATORS &	RESEARCH
& DISCOVERY	Bloomberg 🎲 THOMSON REUTERS D DOW JONES Quand		Z zoominfo acxiem. experian.	FOURSQUARE O mapbox sense360	EATA.GOV Dittation		SCHOOLS	OpenAI facebook research
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Version 1.0 - September 2020

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mattturck.com/data2020

74

The raise of Artificial Intelligence (WARZiupo)

- It is the science and engineering of making intelligent machines.
- In Computer Science, Artificial Intelligence (AI) research is defined as the study of « intelligent agents »
- From General AI to Narrow AI: from overhyping to fewer promises, but more realistic!





Since an early flush of optimism in the 1950's, smaller subsets of artificial intelligence - first machine learning, then deep learning, a subset of machine learning - have created ever larger disruptions.





- General-purpose AI like the robots of science fiction is incredibly hard
 - Human brain appears to have lots of special and general functions, integrated in some amazing way that we really do not understand (yet)
- Special-purpose AI is more doable (nontrivial)
 - E.g., chess/poker playing programs, logistics planning, automated translation, speech and image recognition, web search, data mining, medical diagnosis, keeping a car on the road.







- Proposed By Alan Turing in 1950
- To be called intelligent, a machine must produce responses that are indistinguishable from those of a human.
- Human judge communicates with a human and a machine over text-only channel.
- Both human and machine try to act like a human.
- Judge tries to tell which is which.
- Is Turing Test the right goal?

"Aeronautical engineering texts do not define the goal of their field as making 'machines that fly so exactly like pigeons that they can fool even other pigeons."" [Russell and Norvig]









if AI can be more **rational** than humans in some cases, why not?



Systems that think	Systems that think			
like humans	rationally			
Systems that act	Systems that act			
like humans	<u>rationally</u>			

AI focus on **action**. Avoids philosophical issues such as "is the system conscious" etc.





AI TECHNOLOGIES

STRONG / GENERAL AI	WEAK / SPECIALIZED KI					
	CLASSIC / SYMBOLIC					
	TABLE BASED DECISION TREES AGENTS					
	SEARCH EXPERT SYSTEMS					
	RULE BASED SYSTEMS SYMBOLIC LOGIC					
	EXPERT SYSTEMS DECISION TREE WITH NN INPUT LEARNING					
	MONTE CARLO XGBOOST					
	SEARCH WITH NN MIXED ALGORITHMS					
	NEURAL NETWORKS					
	LSTM NAIVE BAYES					
	CONVOLUTIONAL NN RANDOM FOREST					
	ATTENTION Q LEARNING					
	AUTOENCODER EVOLUTIONARY					
	DEEP LEARNING ALGORTHIMS					
<	MACHINE LEARNING					





- Develops Narrow Artificial Intelligence systems through examples
 - A developer creates a model and then "trains" it by providing it with many examples
 - The machine learning algorithm processes the examples and creates a mathematical representation of the data that can perform prediction and classification tasks

• Example

 A machine-learning algorithm trained on thousands of bank transactions with their outcome (legitimate or fraudulent) will be able to predict if a new bank transaction is fraudulent or not



Machine Learning Techniques





81





ML model is presented with *input data* which is labeled
 Each *input data* is tagged with the correct label.

The goal is to approximate math operations in the ML model so well that when presented with new *input data*, the ML model can <u>predict</u> the output variables for that *input data*.







- On the left side of the image, some data is marked as 'Spam' or 'Not Spam'. This is *labeled data.* This data is used to train the supervised model, the *intelligent* program (at center of the image).
- Trained model is tested with new mails (on the top of the image) and checking if the output of the supervised model is correct (on the right side of the image).







- **Classification**: A classification problem is when the output is a category, such as "red" or "blue" or "disease" and "no disease".
- **Regression**: A regression problem is when the output is a real number, such as "dollars" or "weight".







- Dependent variables: the main event or factor to understand or predict. Also known as *explanatory variable.*
- Independent variables: the events or factors suspected to have an impact on the dependent variable. Also known as *response variable*.





Simple regression: single independent variable for a single dependent variable. It is very common to name the independent variable as *x* and Y as the dependent variable *x*: number of cricket chirps
 Y: temperature

Multivariable regression: multiple

independent variables, x_1 , x_2 , x_{3} , for a dependent variable **Y**.

- x_1 : number of cricket chirps
- x_2 : rainfall
- x_3 : automobile traffic
- Y: temperature







- ML model is presented with unlabeled, uncategorized data
- ML model acts on the data without prior training.
- The output is dependent upon the coded algorithms.
- Is one way of testing AI.







- In the below example, some cartoon characters are passed to the ML model. Some of them are ducks.
- No data label provided.
- ML model is able to separate the characters into 'Duck' and 'No duck' by looking at the type of data and models in the underlying data structure.







- **Clustering**: Discovering the inherent groupings in the data, such as grouping customers by purchasing behavior.
- Association: Discovering rules that describe large portions of the input data, such as people that buy X also tend to buy Y.







Association



- A reinforcement learning algorithm, or agent, learns by interacting with its environment.
- The agent receives rewards by performing correctly and penalties for performing incorrectly.
- The agent learns without intervention from a human by maximizing its reward and minimizing its penalty.
- It is a type of dynamic programming that trains algorithms using a system of reward and punishment.









- The agent is given 2 options i.e. a path with water or a path with fire.
- If agent takes the fire path then a penalty is subtracted
- Agent learns it should avoid the fire paths.
- If agent takes water path then some a reward is granted
- Agent learns what path is safe and what path isn't.



Beyonds Machine Learning?



- Combines advances in computing power and special types of Neural Networks to learn complicated patterns in large amounts of data
- State of the art for identifying objects in images and words in sounds
- Applied successes in pattern recognition to more complex tasks such as automatic language translation, medical diagnoses and numerous other important social and business problems



s, smaller subsets of artificial intelligence - first machine learning, then - have created ever larger disruptions.



Neural Networks: the Perceptron

- Mathematical representation of a biological neuron
- First implementation by Frank Rosenblatt in the 1950s
- Rosenblatt's perceptron is activated when there is sufficient stimuli or input. (Neurons have been found to perform a similar process, in which experience strengthens or weakens dendrites' connections)



How does a Perceptron Work?



- Perceptron receives the value of the attributes of an input, just as dendrites do in a neuron.
- Each attribute has a weight that measures its contribution to the final result, which is the sum of the multiplications of inputs of each attribute by its corresponding weight.
- If the sum is greater than zero Perceptron returns a value of 1, otherwise it yields 0.







- Neurons by themselves are kind of useless, in large groups, they work together to create some serious magic!
- Neural Networks are no more than a **stacking** of multiple *perceptrons* in layers to produce an output.
- Input into one layer that creates an output which in turn becomes the input for the next layer, and so on. This happens until the final output signal.







- In the 1980s, most Artificial Neural Networks (ANN) were single-layered due to the cost of computation and availability of data.
- Nowadays is possible to afford more hidden layers in ANN, hence the moniker "Deep Neural Networks" (DNN).
- Regained popularity since ~2006.
- Rebranded field as Deep Learning (DL)



Types of Deep Neural Networks

• Feedforward Neural Networks (FFNs, ANNs or NNs)

Original

Encoder

- Recurrent Neural Networks (RNNs)
- Convolutional Neural Networks (CNNs)
- Autoencoder Neural Networks (AEs)



Decoder











(«WAZiup») («WAZihub»)

- Contain five types of layers:
 - Input
 - Convolution
 - Pooling
 - Fully connected
 - Output.



- Each layer has a specific purpose, like summarizing, connecting or activating.
- CNN have popularized image classification and object detection.
- Also applied to other areas, such as natural language processing and forecasting.

Machine/Deep Learning for scient

- Large variety of supported languages
 - Python, R, C++, Java, Scala, Javascript, Go, ...
- Many statistical methods/algorithms are implemented in libraries
- Examples
 - Scikit-learn
 - Google TensorFlow
 - Microsoft Distributed Machine Learning Toolkit
 - Apache Mahout
 - ...
- But, beware
 - There are hundredth of tools...
 - ...and new tools every months!







• https://playground.tensorflow.org/

Tinker With a Neural Network Right Here in Your Browser. Don't Worry, You Can't Break It. We Promise. Epoch Learning rate Activation Regularization Regularization rate Problem type 3 000,000 0.03 Tanh 0 Classification None 2 HIDDEN LAYERS DATA FEATURES + -OUTPUT Which dataset do Which properties do Test loss 0.522 you want to use? you want to feed in? Training loss 0.508 + -+ -4 neurons 2 neurons -X₁ X2 Ratio of training to test data: 50% X1² The outputs are mixed with varying weights, shown Noise: 0 X2² by the thickness of the lines. Batch size: 10 X₁X₂ This is the output from one neuron -6 -5 -4 -3 -2 -1 **0** 1 2 3 4 5 6 Hover to see it larger sin(X₁) REGENERATE Colors shows data, neuron and sin(X₂) weight values.

101

INTEL-IRRIS

OBJECTIVES

PILOTS CONSORTIUM

IOT+AI ILLUSTRATION

INTEL-IRRIS

Intelligent Irrigation System for Low-cost Autonomous Water Control in Small-scale Agriculture



Intel-Irris









Intel-IrriS June 2021 – May 2024

Intelligent Irrigation System for Low-cost Autonomous Water Control in Small-scale Agriculture

> Propose low cost but highly efficient water control systems for irrigation optimization

> Use cutting-edge technologies to propose highly innovative systems yet simple to deploy and adapted to smallholders

Seamless integration into existing irrigation system and/or local customs and practices

Improve farmer's knowledge on water-related issues, foster local adaptation of technologies, increase local innovation capacity and facilitate technology appropriation

Large-scale adoption of low cost smart irrigation system by smallholders, stimulating synergies between various local actors



Prof. Congduc Pham http://www.univ-pau.fr/~cpham



Low-cost sensors: accuracy?





- Build on low-cost, low-power IoT expertise
- Increase accuracy of low-cost sensors by advanced calibration
- Enable deployment of several complementary low-cost sensors: soil conductivity, volumetric water content, ...
- Include agricultural models / knowledge with corrective & predictive analytics



Understanding soil water



Source: Christian Hartmann, IRD

• Low-cost sensors usually measure soil water content

• Soil = a pile of aggregates 23 phases: solid + air & water







Source: Christian Hartmann, IRD

in the soil, the water is UNDER TENSION
 = it is hold by CAPILLARY FORCES

Water tension is also needed!









Smart embedded control



- Build on low-cost embedded
 & open IoT gateway expertise
- Implement the "Intelligent Irrigation in-the-box" vision
- Model complex interactions: water-soil-plant interaction, evapotranspiration,...
- Embed Decision Support System (DSS) and disruptive Artificial Intelligence (AI)
- Integration of multiple knowledge streams
- Fully autonomous



Edge-Al for fully autonomous system inub

 Embed every thing on the IoT gateway to provide a fully autonomous system for the "Intelligent Irrigation-in-the-box"




Edge-Al integration









- Internet-of-Things provides the unique feature to make things "talk" to us: localisation, surrounding environmental conditions, particular events, ...
- It has huge potential into helping humanity to reach the UN SDG (Sustainable Development Goals)
- With mode complex sensors such as cameras, spectrometers, hyperspectral cameras,... we can get better knowledge to further optimize a number of process for sustainable development
- LIUPPA works for more than
 8 years to develop & deploy low-cost IoT in Africa with 4 EU H2020/PRIMA projects



IOT Online Course

Fundamentals of IoT

Continue with F-IOT-2c: Introduction to IoT hardware

