How Artificial Intelligence Can Help In The Fight Against Climate Change

### AI &COGNITIVE SYSTEMS FORUM

CONNECTED TRANSPORT

**MANUFACTURING** 

**HEALTHCARE** 

**ENERGY AND UTILITIES** 

**BUILDINGS & INFRASTRUCTURE** 

**OPEN INDUSTRY** 

**ENABLING IoT** 







# Alicante Context (Spain)









1930s



Touristic
Urbanization
1950-70



1980



2018



# Aguas de Alicante

1898 \_\_\_\_\_\_\_ 1953





(\*) First Mixed Capital Company of the water sector in the world (according to World Bank)



## City challenges



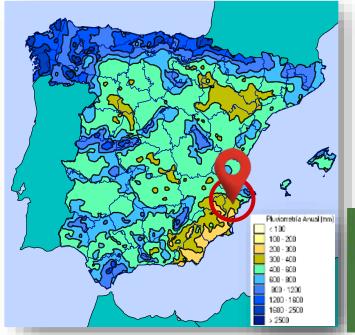


### Alicante faces complex conditions

- Semi-arid climate
- No local water sources
- Seasonal torrential rains
- Impact of tourism

### Responsible and efficient management

- Water resources preservation
- Environmental care
- Preventive strategy for urban resilience









# Research, Development and Innovation R&D Projects [1/3]











#### **SIPAID**

- Flood and CSO early warning system
- Online access for the Municipality and Emergency Services
- Short and médium term forecast, radar, rain gauges and level gauges

#### Augmented Reality for Street works

- Aimed at citizen communication
- Real time info on work planning, progress, street closures..
- Web publication + AR app

#### **ICAP**

- Regional platform for water quality monitoring and warnings
- Jointly with Mancomunidad de Canales del Taibilla (Public bulk wáter provider)

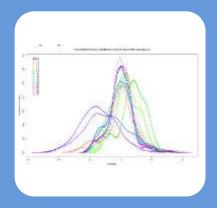
#### Smart Irrigation

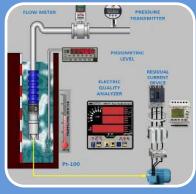
- Urban irrigation optimization
- Analyses plant species, rain, weather forecast...
- Industrialised by Advanced Solutions



# Research, Development and Innovation R&D Projects (2/3)











#### **PALACE**

- Multi-model software for water demand prediction, based on a hybrid (time series + external factors) approach
- Short time: 1-6 days
- Long time: 1 year

#### **IDROSMARTWELL**

- Patented level sensor for groundwater wells
- Real time control of hydrogeological + technical parameters
- Optimization of energy consumption and maintenance

# RECYCLING OF EXCAVATION MATERIALS

- Processing of excavation materials for trench backfilling
- Technical suitability tests for validation
- Economic and environmental analysis

#### DAIAD

- European FP7 project
- IoT device and app for water and energy consumption monitoring
- Winner of the EC Innovation Radar Prize (Tech for Society Award)



### Research, Development and Innovation AGUAS R&D Projects (3/3)











INLOC\*

**SOUNDWATER\*** 

**ROBOTOUR\*** 

**CIS-WATER** 







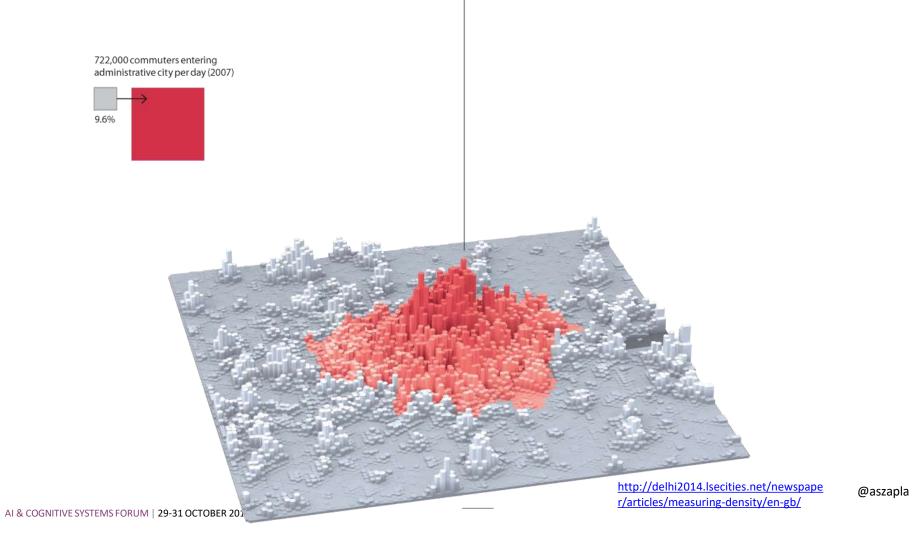


https://www.youtube.com/watch?v=2WGPvWPpey8



#### **LONDON**

### #urbanfooprint

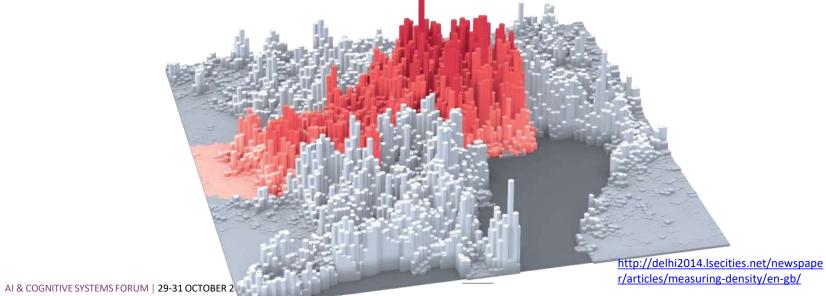


21,000 people/km<sup>2</sup>



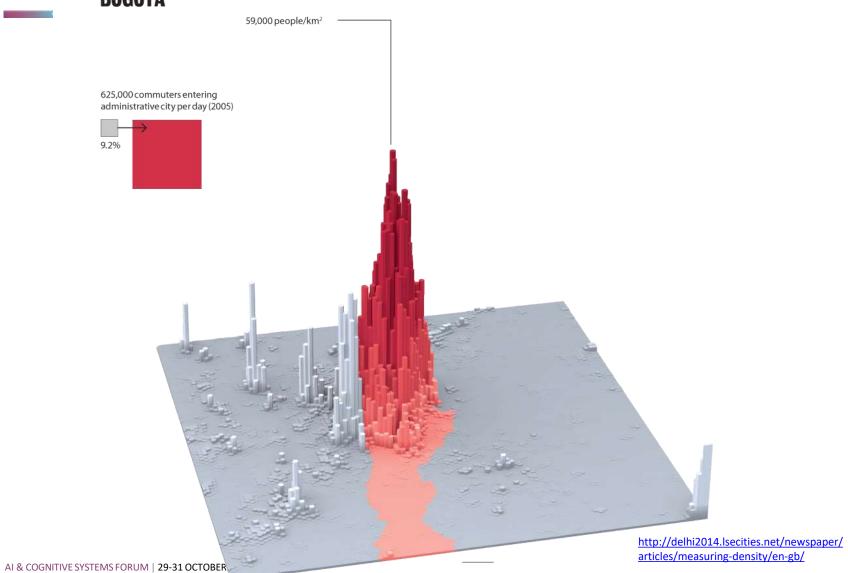
2,530,000 commuters entering administrative city per day (2010)

20%



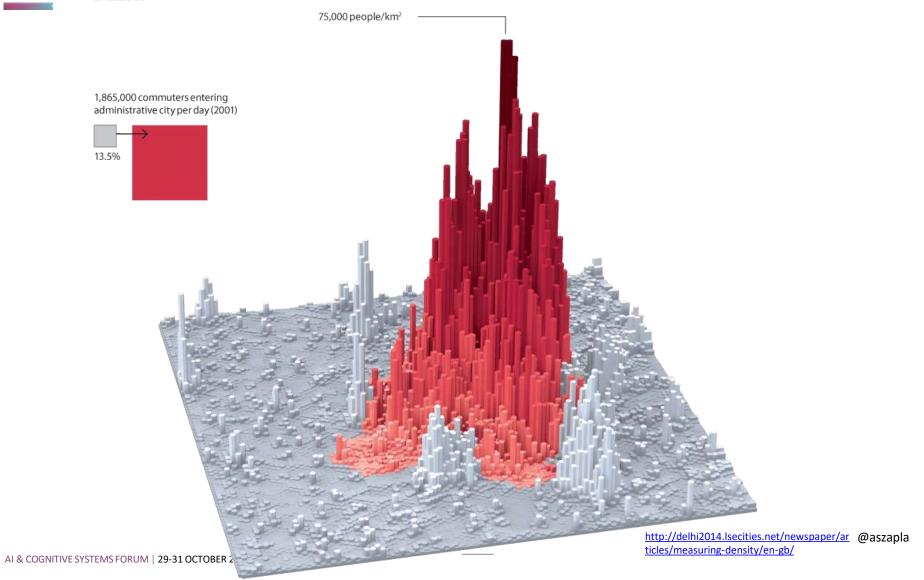


#### BOGOTÁ

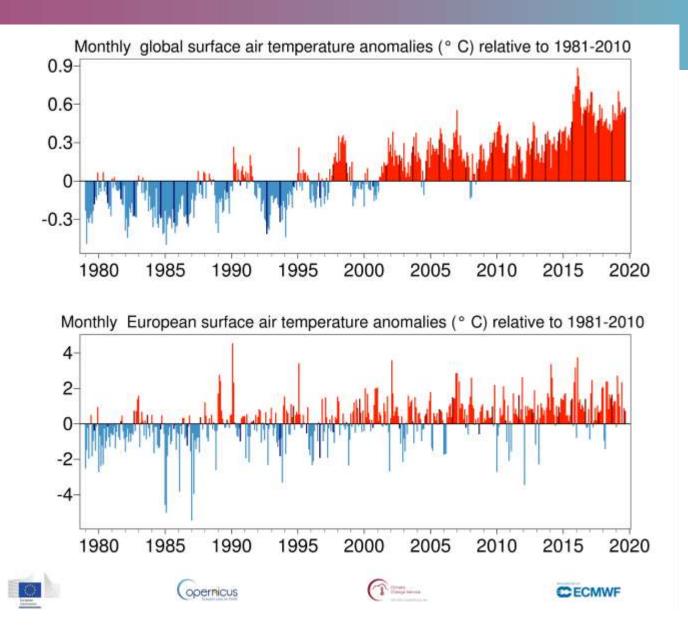




#### DELHI

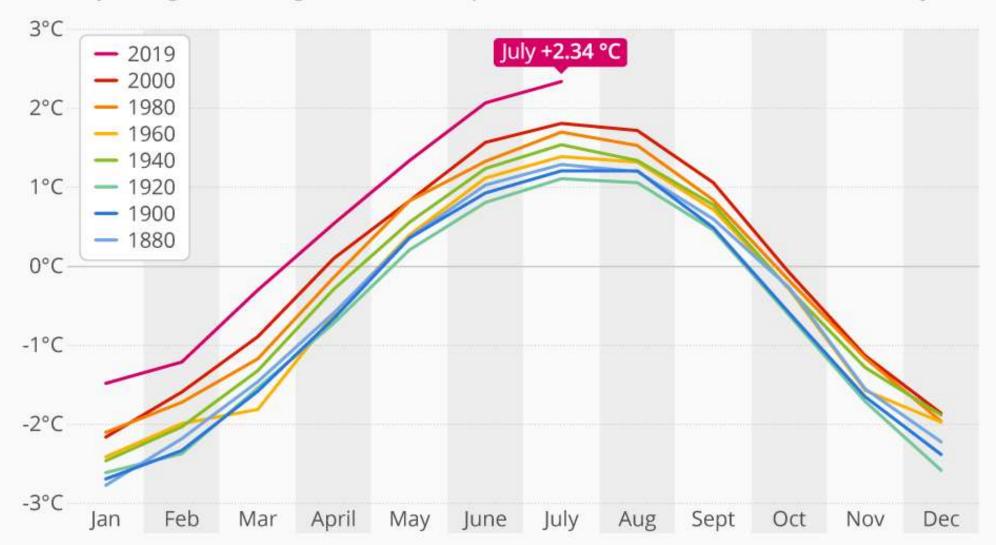






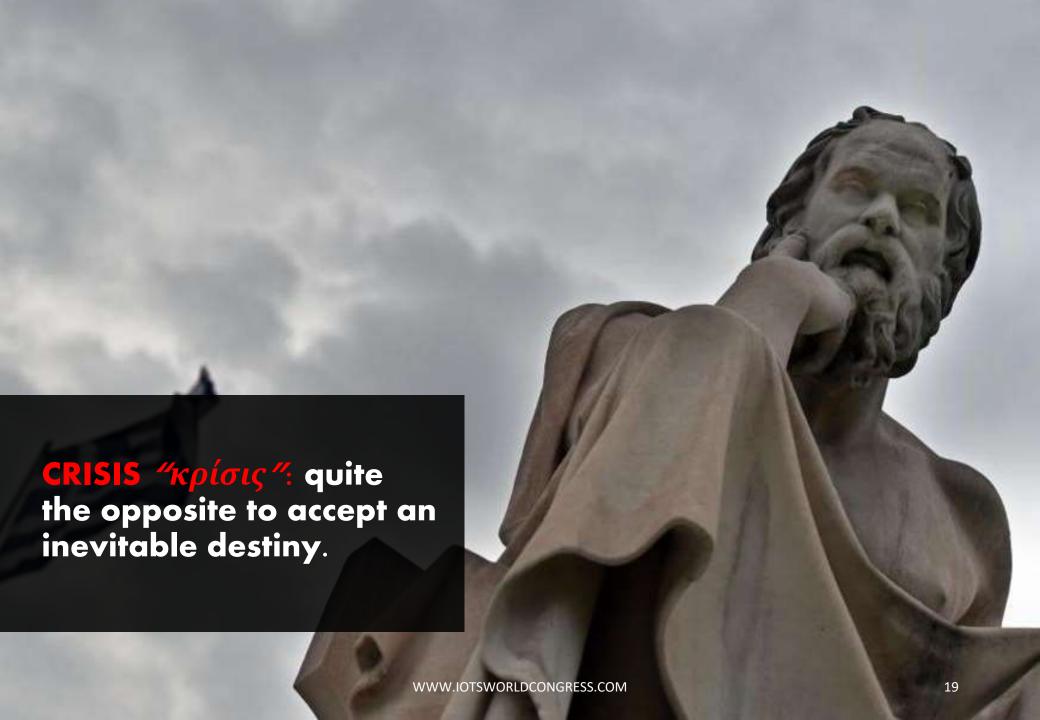
### **Earth is Heating Up**

Monthly divergence from global mean temperature (between 1980-2015), in selected years









# 12 Sept 2019

# Diez medios áereos y 43 bomberos trabajan en un incendio forestal en Paterna del Campo (Huelva)

Actualizado 12/09/2019 16:07:20 CET



#### Últimas noticias / Andalucía >>

- Andalucía reconoce la excelencia de profesionales y empresas que impulsan la industria turística del destino
- El sorteo de Euromillones deja 2,5 millones de euros en Dos Hermanas (Sevilla)
- A prisión los dos detenidos al intentar alunizar una tienda de Lagoh en la que no lograron robar

#### **FIRES**

 Hospital de Dia del Infanta Elena reducirá las complicaciones quirúrgicas



# Agua en Ávila solo hasta la primera semana de noviembre

#### **DROUGHTS**



Sequía en Ávila / RAÚL HERNÁNDEZ (ARCHIVO)

El consistorio ha anunciado que las reservas de agua, hoy al 29.8%, solo se garantizan hasta esa fecha, si no llueve o hay alguna incidencia



### **FLOODS**



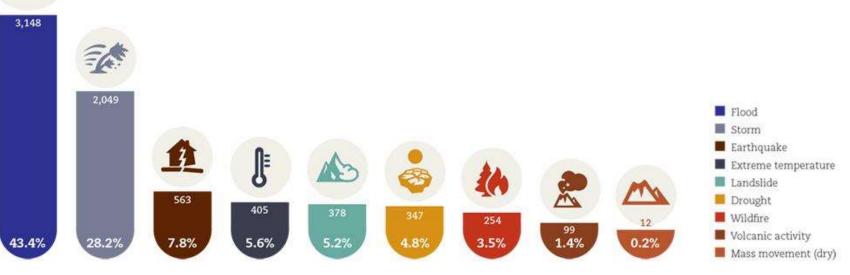




### Number of disasters by type - 1998-2017

13 CLIMATE









RR #GlobalGoals



# SUSTAINABLE GEALS DEVELOPMENT GEALS



















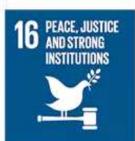
















# 2. Al vs SGD



### 10+1 Key concepts about Forecasting

- 1. Determine the use of the forecast;
- 2. Choice the forecast horizon and forecast approach;
- 3. Collection and analysis of data;
- 4. Identification of the forecast model(s);
- 5. Estimation of the forecast model(s);
- 6. Diagnosis of the statistical adequacy of the model(s);
- 7. Production of the forecast, including confidence intervals;
- 8. Evaluation of the forecast;
- 9. Use of the forecast by decision makers;
- 10. Ex-post facto analysis of forecast error.

And.....

### 0. Asking the Right Questions



# 4 steps to "knowledge cake"



## **FATEN**

F: Fairness

A: Autonomy,
Accountability and intelligence
Augmentation

T: Trust and Transparency

E: Education, bEneficence and Equality

N: Non-maleficence





## Al vs DROUGHTS:

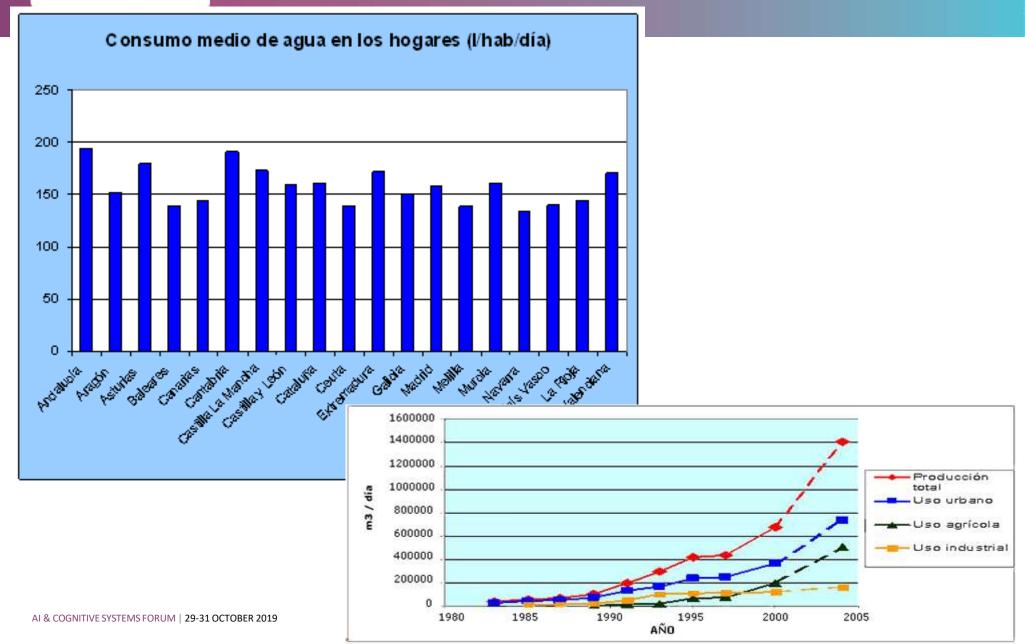
# WATER DEMAND FORECAST SOLUTIONS





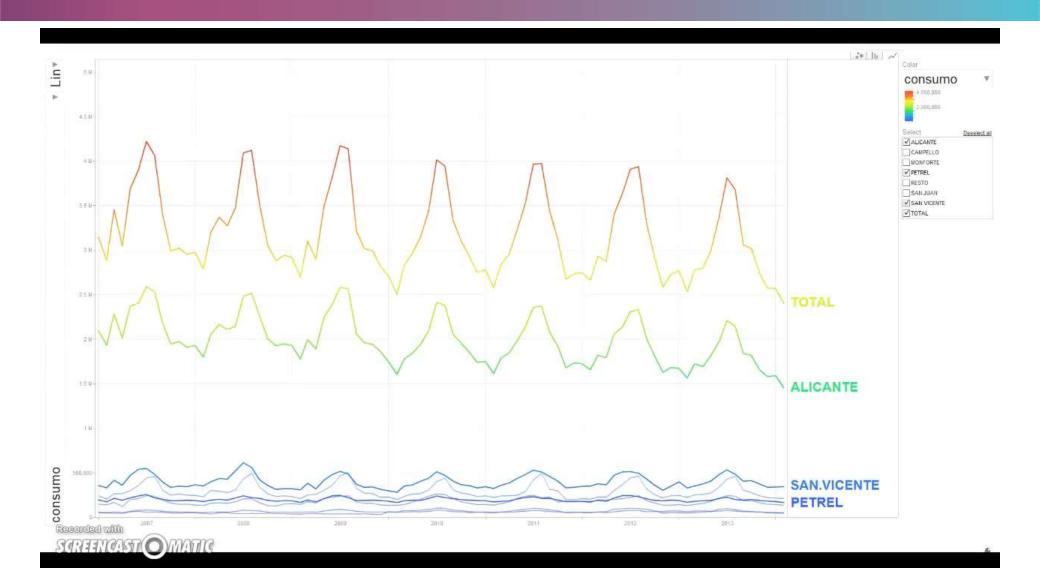


## "Traditional" way Water Demand





# New ways









### **Alicante Water Demand Forecast Solutions**



#### **CITY LEVEL**

1. Short Term WD

2013

2. Medium Term WD



### **CUSTOMER LEVEL**

3. Medium Term Revenues

2015



#### DOMESTIC LEVEL

4. Short Term WD

2016/17

AI & COGNITIVE SYSTEMS FORUM | 29-31 OCTOBER 2019



# City Level

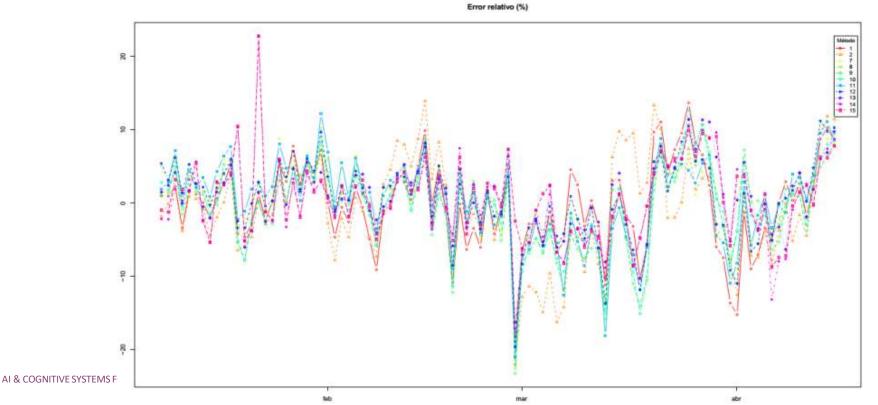
### How to reduce operations and purchasing water costs?

- SHORT TERM (1 to 6 days)
  - Supply Forecast required for Operations Management
  - Previously solved through Excel Spreadsheet tool
- MEDIUM TERM (1 to 18 months)
  - Supply Forecast required for Water Purchase decision making
  - Key for Strategic Planning



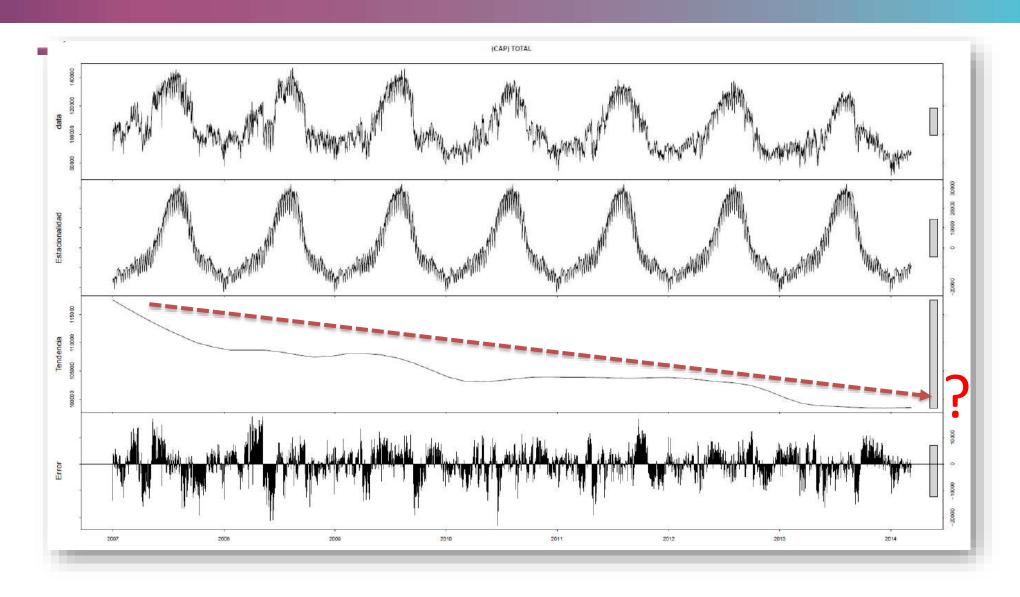
### **Project Description**

- 13 Short Term methods and 6 Long Term methods tested.
- Methods selected for their minimum error / shortest computation time:
  - Short term: Method based on holiday calendar and 6 days forecast of temperature/rainfall:
    - Alicante, San Vicente, San Juan, Monforte, Petrer: Method No. 13
    - Campello: Method No. 7 (fitted to greater seasonality)
  - Long term: Method based on time series for six months projection.





### Seasonal Breakdown of trends (1 year daily mean)





# SHORT TERM FORECAST SYSTEM

# **Contracts consumption**

Alicante, Campello, Monforte, San Juan, Petrel, San Vicente, Rest (and total consumption).

From 7 days to 6 years back

Statistical Adjustment and Forecast for each request.
6 Days Projection

### **Holidays Info**

Holidays according to the associated target grouping

# Weather forecast

Daily prediction for the next 6 days.

Minimum, mean and maximum temp.

Defined for each target

Forecast Module

> Daily Automatic Execution

### **WD WEB**

Configuration

Presentation of results



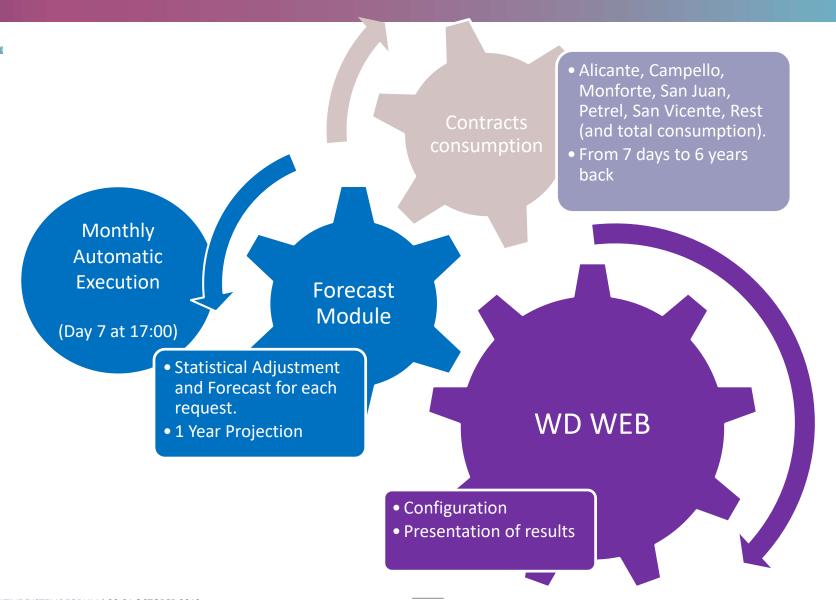
### SHORT TERM FORECAST AUTOMATIC REPORT

#### **REPORTS**





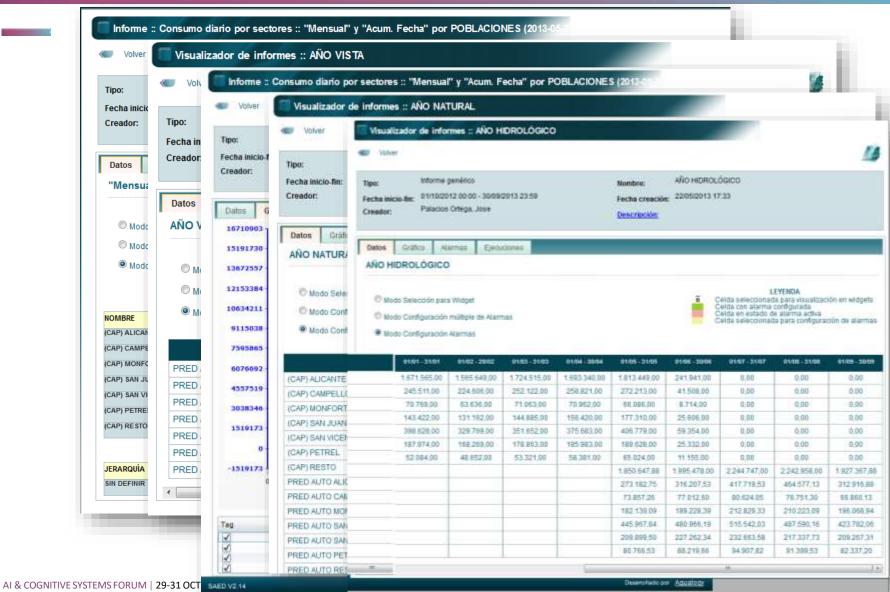
# MEDIUM TERM FORECAST SYSTEM



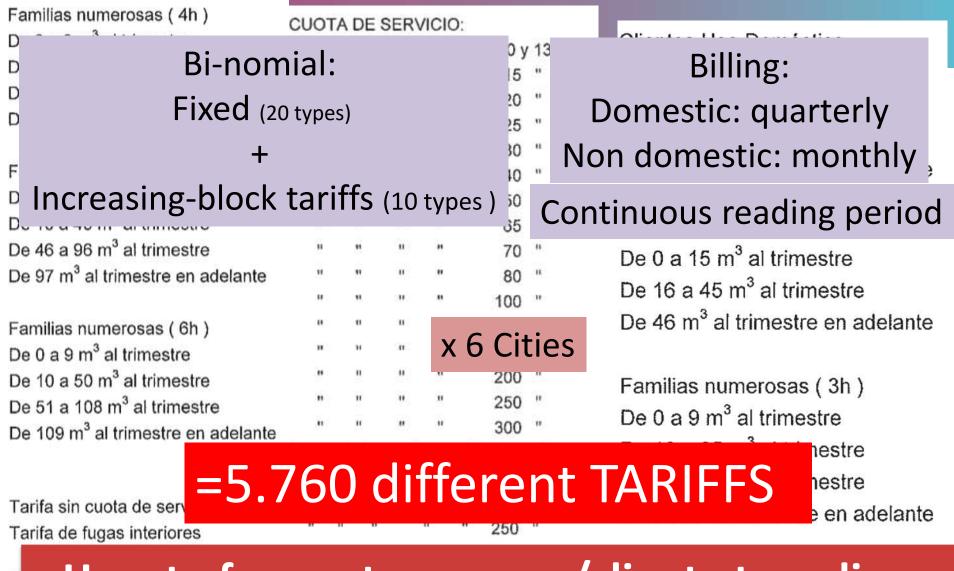


### MEDIUM TERM FORECAST AUTOMATIC REPORT

#### REPORTS



# Customer Level



How to forecast revenues/client at medium term (up to 2 years)?



# **Customer Level**

# **SOLUTION 1: Clustering**



- Contract profiling: Aggregation of contracts into homogeneous profile (urban, rural, touristic, etc.)
- Build a new panel of customer
- Selection of best model for each segment
- Calibration of best model for each segment
- Forecast -> average tariff

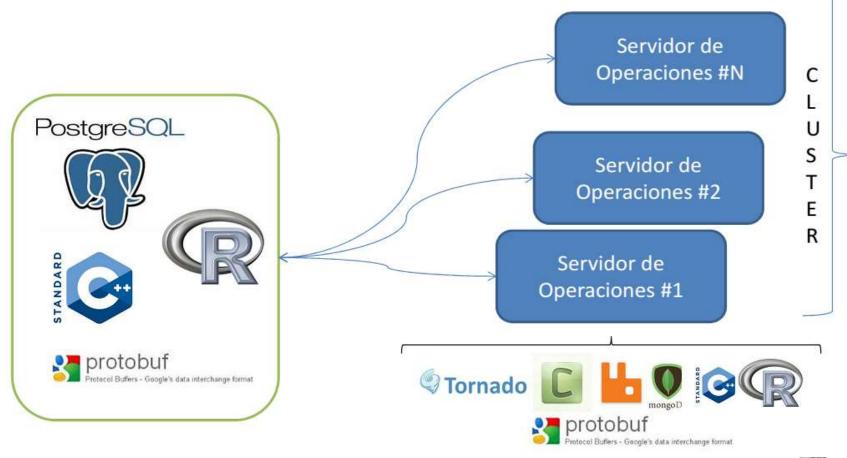
# **SOLUTION 2: Per client**



- 7 Gb data to be processed
- Selection of best model for each client!
- Calibration of best model for each segment

308.322 series

Alocos Forecast 22-> oread tariff









#### HP Proliam SL 390 G7 (1x2U, 24x1U)

- Memoria 46 GB DDR3-1333



# **Domestic Level: DAIAD**

### How to forecast customer hourly consumption?

#### No idea about actual water use!

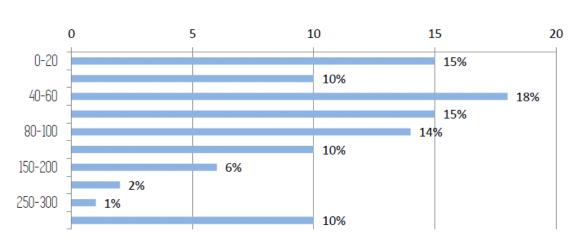


Figure 2: Responses on "how much water do you consume every day? (in liters)"



**Emphasis on the shower (2<sup>nd</sup> largest energy use)** 













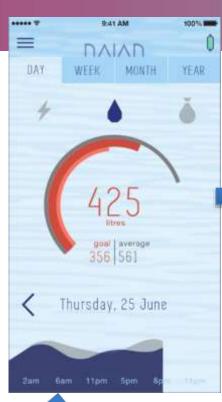


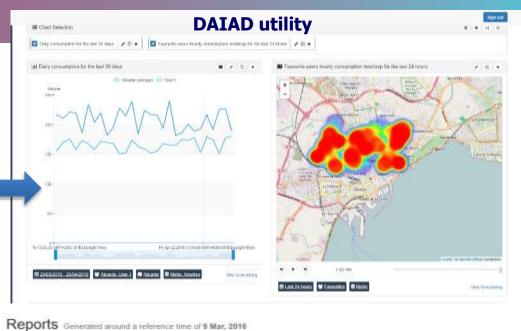




# **DAIAD Project**







**DAIAD Device** 

**DAIAD APP** 

### **AMR System**





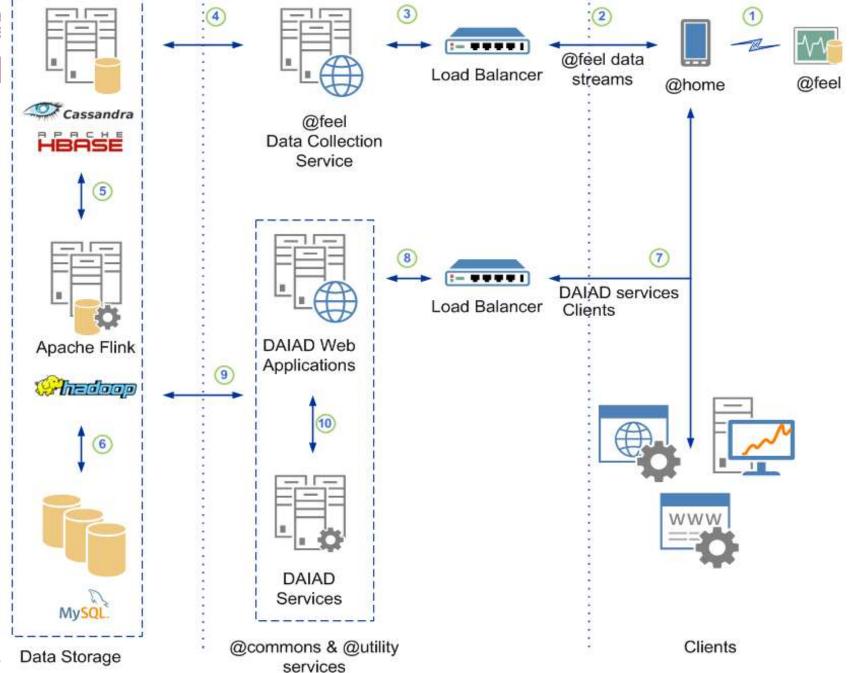
Use reference time: 9 Mar, 2016

Total weekly consumption - Last Week

Meter (SWM) + Volume









# **DAIAD** Big Data

Name Category Short Description

Consumers Water Consumption Classification - Regression

Computes and predicts
consumers' water
consumption (in liters) for
the preferable time intervals

**Consumers Clusters** 

Clustering

Finds groups of similar consumers based on their water consumption for the preferable time intervals











#### Machine learning in water network assets management



**Drinking Water Supply System Assets** 

6 drinking water treatment plants

77 water tanks

84 pumping stations

4,700 km of network

Network asset management challenges:

- 1. How the 130.000 pipe sections will work in the future?
- 2. How minimize the consequences of network failures?



"Unfolding" the network: all pipe sections in a row . . .





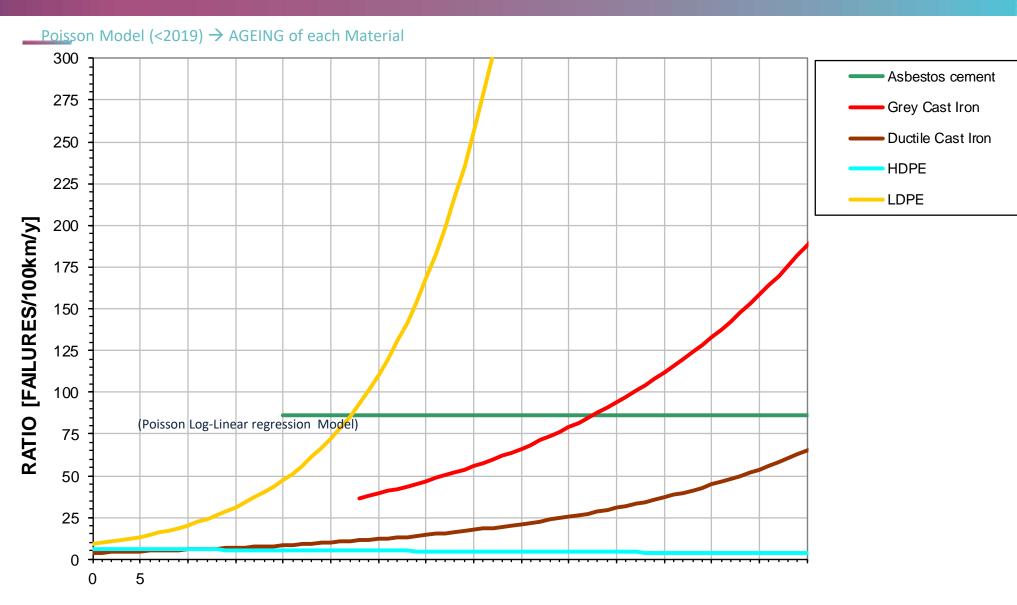
# Water network assets management



Our commitment with sustainability leaded us to use Machine Learning for improving short term failure potential



# Water network assets management

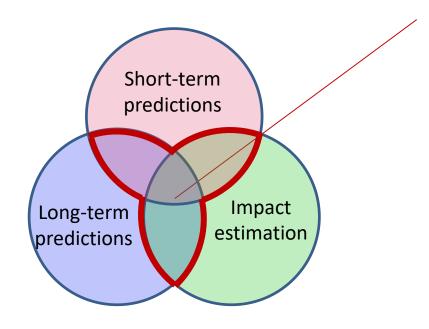


Pipe age [y]





### **Future improvements**



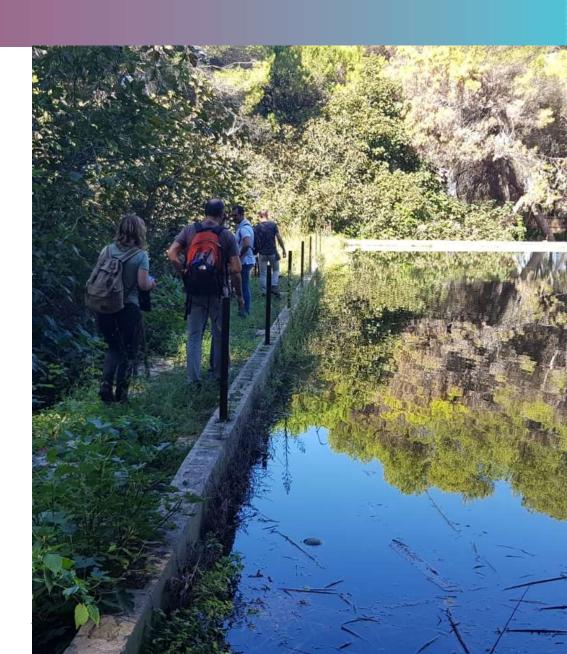
Decisions are made inside this space



# AI vs FIRES: GUARDIAN









# **GUARDIAN Project: Context**



Climate change has increased the fire risk and the threat to the so-called Wildland-Urban Interface

In the Mediterranean context of water scarcity, the European **Urban Innovative Action GUARDIAN** is implementing a fire defense solution based on the use of reclaimed water

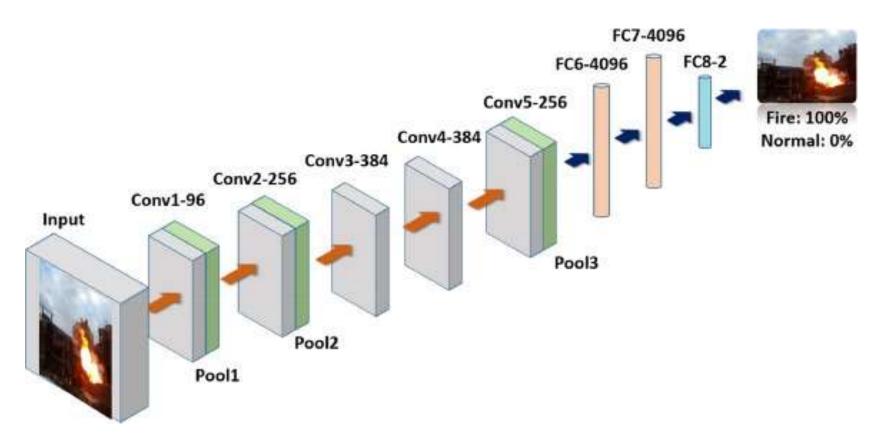


### **GUARDIAN Solution**





### **Deep Convolutional Neural Networks for Fire Detection**

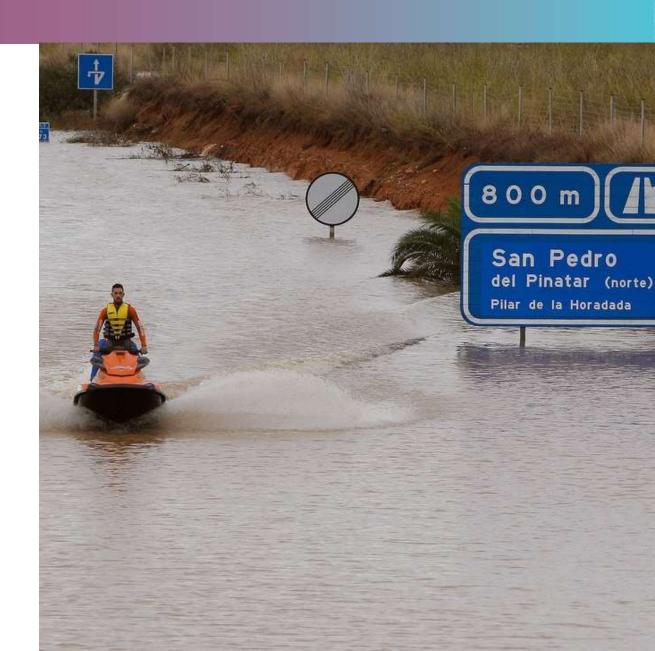


https://www.sciencedirect.com/science/article/pii/S0925231217319203

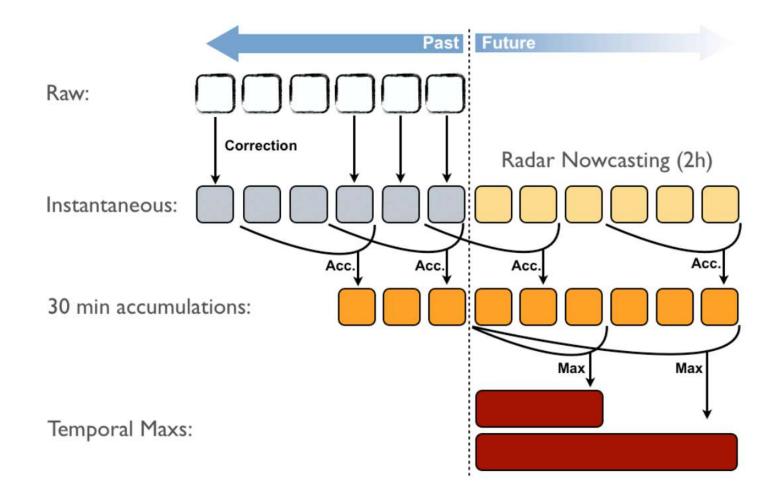


# Al vs FLOODS: FLOODALERT

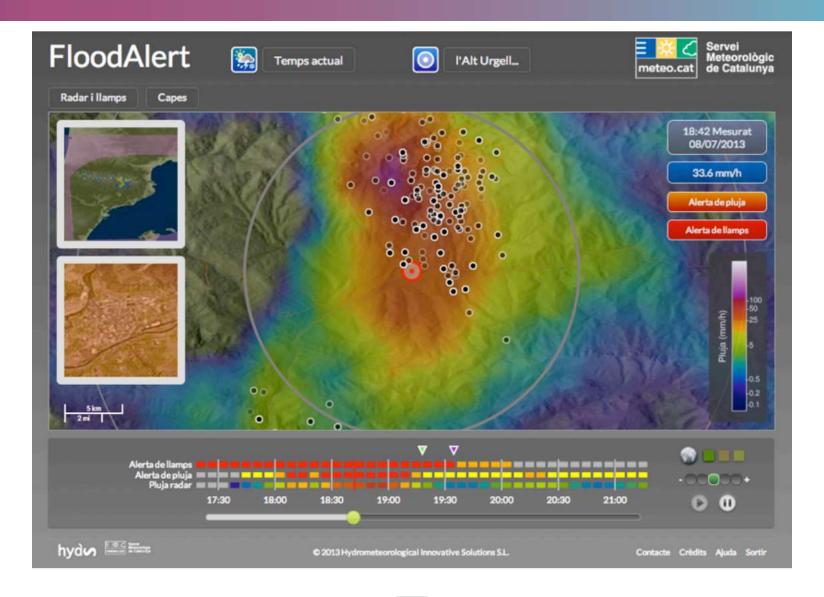




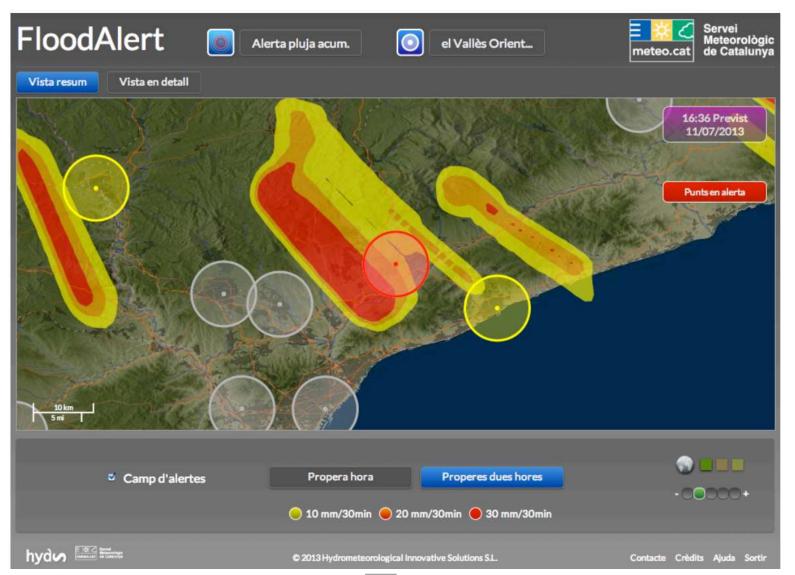




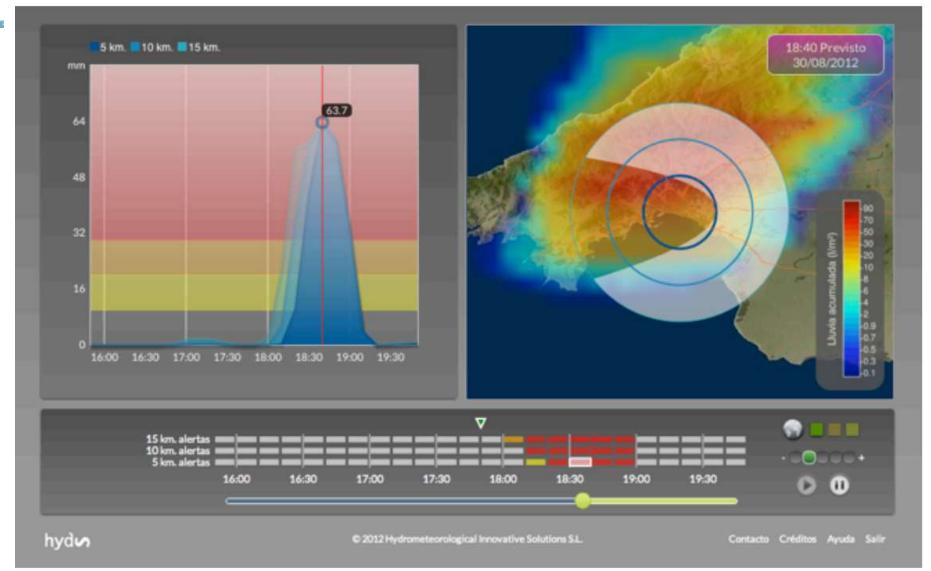


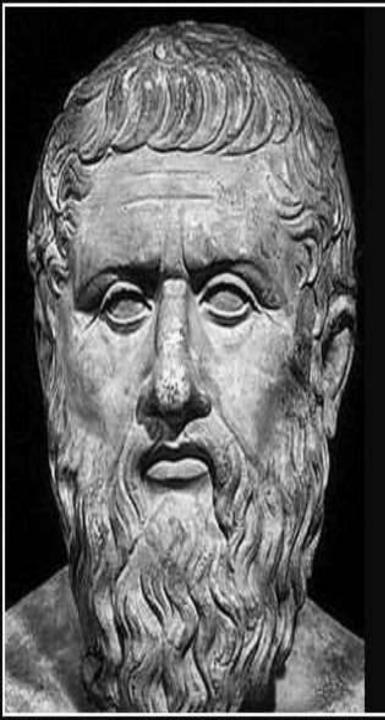












A good decision is based on knowledge, and not on numbers.

— Plato —

**Conclusion** 





















/antonio-sánchez-zaplana/

www.urbaninnovation.es

antonio.sanchez@aguasdealicante.es



# DIGITALIZING INDUSTRIES

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