

# ARTIFICIAL INTELLIGENCE (AI) FOR TERRITORIAL MANAGEMENT

Use cases show principles and benefits of AI for optimizing the management of environmental infrastructures: energy, water, mobility, waste.















# Atos

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We offer our clients a range of state-of-the-art digital solutions and products as well as consulting, cybersecurity and decarbonization services.















# ARTIFICIAL INTELLIGENCE (AI): AREAS OF APPLICATION

#### **Data science**

Data analysis, prediction and decision





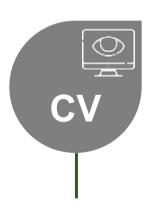
Natural language processing

Language comprehension, human interaction

#### **Robotics**

Interaction with environment and sensors





**Computer vision** 

Detect and extract information from images

# ARTIFICIAL INTELLIGENCE (AI): APPROACHES

### **Supervised learning**

(Input, Output)





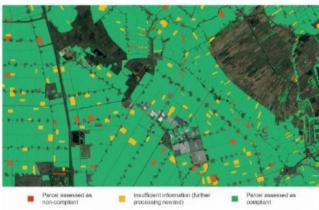


### **Predict**

Classification: is it A or B?

Regression: How much? What will be the value?

### Plots of land classification



### Non supervised learning



## Identify

Clustering: Which organisation?

Association: Who is close?

Outlier detection: who is different?

#### **Fraud detection**



### **Reinforcement learning** Reward







## Learn from mistakes

Decision: What do I have to do now?

Recommendation: What can I offer you?

# USE CASE PRESENTATION: AI FOR THE ENVIRONMENT

- Energy: digital twins of wind farms
- Water: prediction of ground waters level
- Water: prevention of environmental pollution
- Mobility: IOT platform
- Waste: detection of public disorders

## **ENERGY: PREDICTIVE MAINTENANCE**

### Digital twins of wind farms

### **Context**

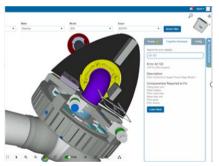


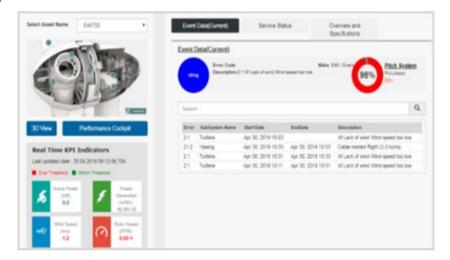
PARK OF 10 FARMS

83 WIND TURBINES (3 TYPES)

2 YEARS SCADA HISTORY

107 SENSORS PER TURBINE





### **Challenges**

It is complex to prevent the appearance of a malfunction. In most cases the flaws are handled at the most critical moment, which leads to important expenses and a substantial shutdown duration of the wind turbine.

This is related to the **lack of knowledge** on the evolution of the deterioration that leads to the breakdown: where does it come from? And when to activate a maintenance operation?

### **Approach**

Proposal of anomaly detection based on prediction error:

1/ The model is trained only on healthy asset data (data collected long before/after the anomalies):



2/ The systel analyses the discrepency towards a healthy asset prediction to identify the risky behaviors:



# **WATER: THE ENKI PROJECT**

### Prediction of ground waters level

### **Context**

Climate change threatens our water resources and their uses (food, agriculture...). Forward-looking management has become necessary for our territories.

### **Approach**

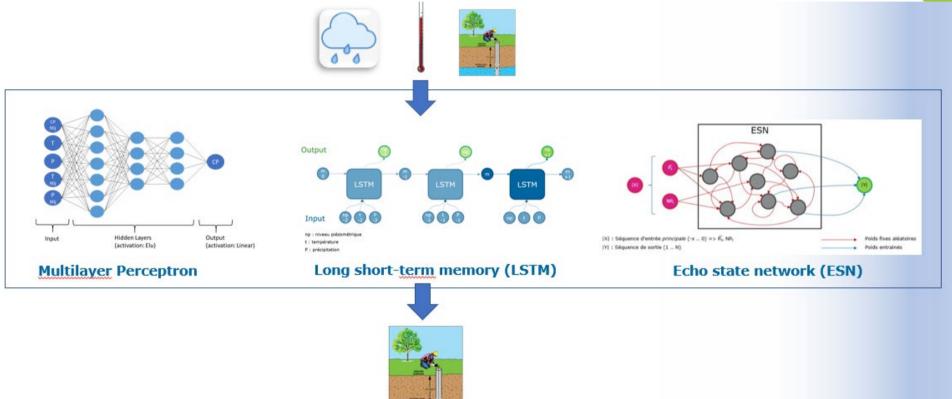
A scoping phase of the project in agile methodology made it possible to identify the needs of twenty partners.

A roadmap has been established to develop a predictive management tool.

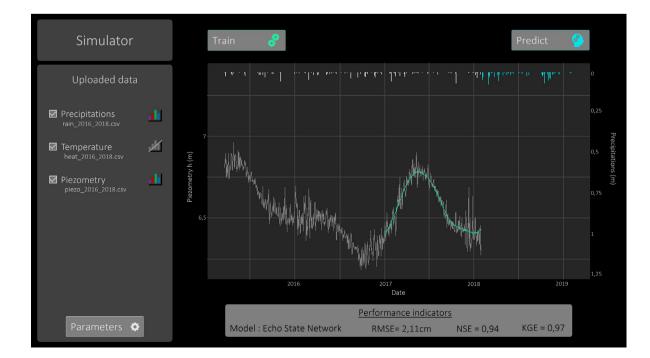




**Machine learning models** to predict piezometry in the short and medium term, based on **climate models**.



Prototype developped during the project







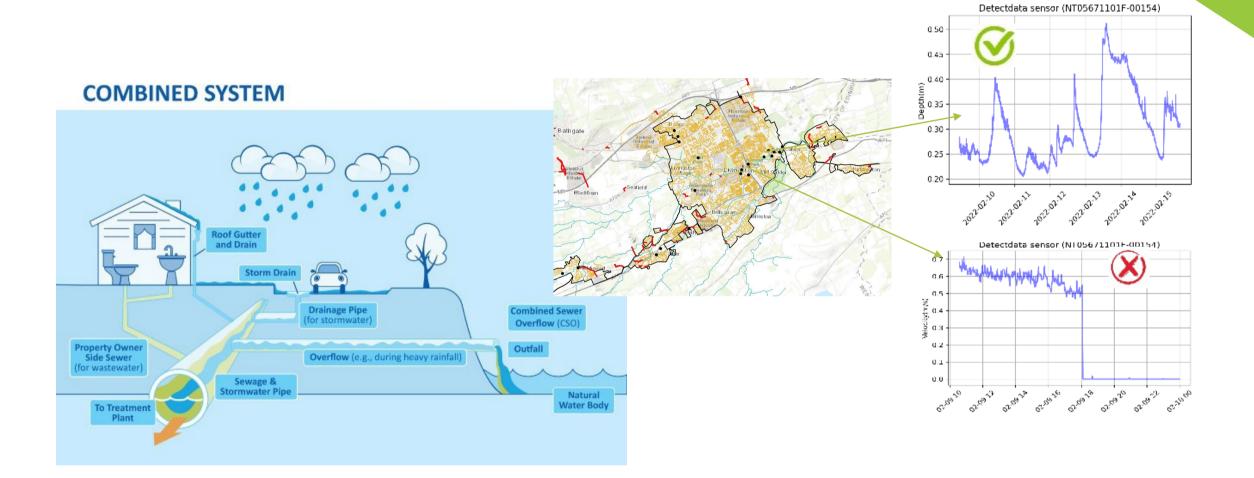
# THE SCOTTISH WATER PROJECT

## Prevention of environmental pollution

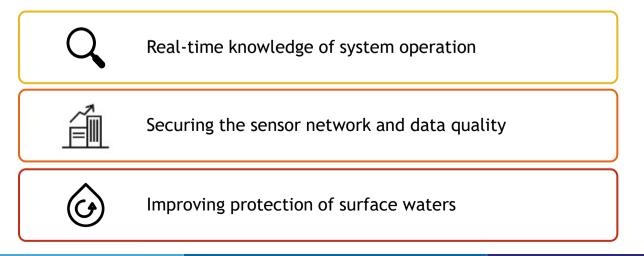


### **Context**

- Wastewater systems can cause environmental pollution or surface flooding.
- For this, an IoT platform has been set up to monitor the wastewater network in real time.
- Machine learning algorithms are able to validate the data quality and to predict levels in the network.



### **Challenges**



### **Approach**

Step 1: Data collection and cleaning

**Step 2**: Tests of different algorithms as needed (Recursive PCA, MLPNN, RBFNN or random forest approaches)

Step 3: Towards a solution in production

# **MOBILITY: IOT PLATFORM**

### A Fiware platform serving a Smart City

### **Context**

With a view to interoperability and standardization, Montpellier Méditerranée Métropole wanted to integrate the FIWARE platform into the Intelligent City information system.

The objective of this Smart City platform is to **ingest and standardize data from IoT and other sensors**.

#### Results

Provision of an Open Data API portal publicly available.

Implementation of data format validation pipeline.

Set up a platform to monitor services in production (alerts, reports).

Implementation of a comprehensive DevOps approach with Continuous Integration and Deployment (CI/CD) automation chains.



### **Solution**

#### The first cases of use concern:

- Management of underground parking lots
- The prediction of free disabled people parking lots
- Optimisation of bicycle parking stations

The architecture allows to easily process data with Al solutions













# **WASTE: THE FIDAMIA PROJECT**

### Detection of public disorders

# AiX MARSEILLE PROVENCE

#### **Context**

To enable all citizens to report public disorder, the FIDAMIA project consists in automating the detection of the type of disorder taken in photo through AI.

The project must have an operational vision from the start and allow for regular updating of the model and possible typologies.

### **Results**

Greater than 80% Al model accuracy commitment across more than 30 classes to be identified.

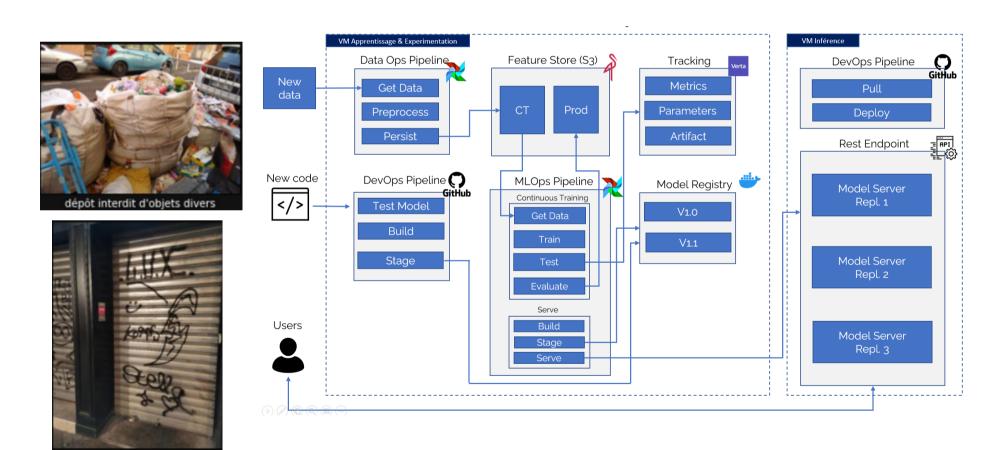
Processing requests in less than 1 second (image processing and inference).

Permanent and automated relearning using MLOps architecture to manage continuous data flow from an initial volume of 500,000 images.

### **Approach**

MLOps Architecture to enable **industrialization and automation** of the entire data science pipeline: data preprocessing, training, non-regression testing.

Deployment of computer vision models:









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### Contacts:

antoine.olgiati@atos.net

loic.maisonnasse@atos.net











