

Shaping Sustainable, Innovative and Secure Smart Cities

22-23 May, 2025 Rabat Smart City, Morocco

#FIWARESummit25

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Shaping Sustainable, **Innovative and Secure Smart Cities**

ML, MLOps & FIWARE

Integrating FIWARE and ML solutions



David Campo

Ph.D. in Computational Intelligence | Sr FIWARE Expert & evangelist

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Agenda

- 1. Learning goals
- 2. Challenges and motivation
- 3. ML and MLOps
 - a. How does ML differ from traditional SW development?
 - b. Towards MLOps: first Reference Architecture
 - c. Approach in FIWARE
- 4. Summary, resources, exercises & QA











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Learning Goals

- ML cycle: differences with typical SW development
- MLOps: what is it?
 - The MLOps life cycle
 - Integration in FIWARE
- Implement in FIWARE solutions
 - Definition of SDM
 - Pipeline for deploying and performing inference











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Challenges and motivation

Motivation:

- Context-driven automation: use the Context Broker to kick different stages of the MLOps process
- Unified context and Observability: the metadata of ML and its operation live in the same context graph as the data, gaining explainability
- Interoperability and Reusability: any ML artifact can be discovered and reused
- Standardized Metadata via SDM: Anyone and any service will exactly know what model.version is or how to interpret job. status
- Event-based MLOps pipelines: the flow of actions is naturally triggered by the nature of the FIWARE system, guided by subscription/notification mechanism
- Decoupled microservised-friendly design: each ML and MLOps subsystem subscribes to specific NGSI-LD entities, making the system easy to change when needed











Challenges and motivation

Challenges:

- Data modelling complexity: this SDM definition can be non-trivial and as complex and granular as wanted. Important to manage the level of detail
- Trigger-Orchestration semantic and complexity: extra thing to take into account, ensuring that subscriptions are properly triggered, with the proper throttling and avoiding race conditions in case of chained multiple subscriptions
- State managing and versioning could fastly became chaotic and excessive complex
- Security and Data Governance: when it comes to sensitive data, as ML models, need to keep an eye to security, adding a point of complexity







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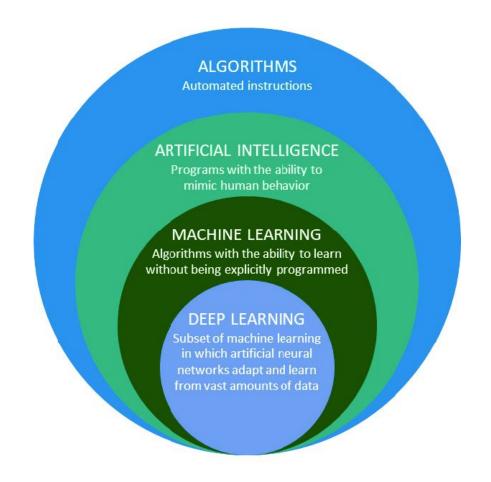


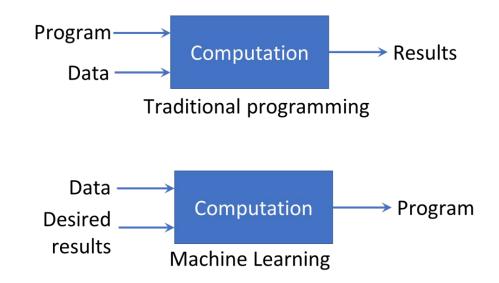






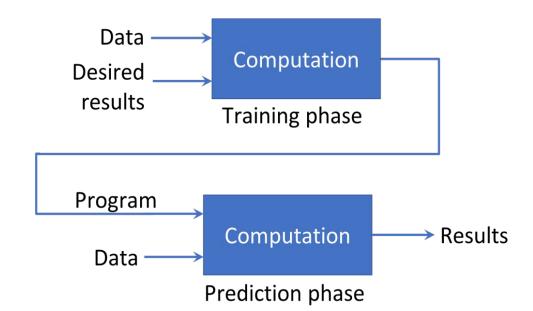
ML and MLOps





The simplest machine learning pipeline is:

- train data and produce a model
- deploy the model and make inference





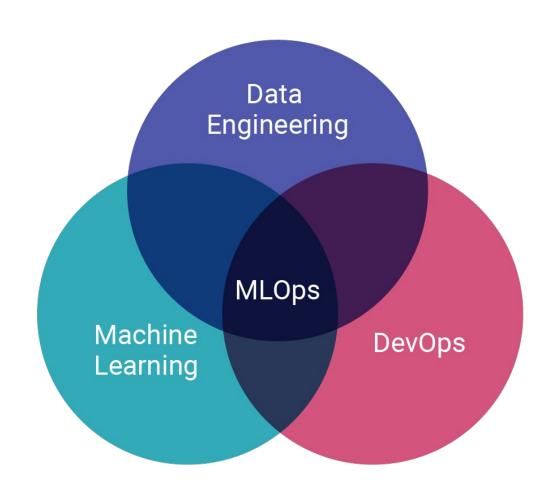


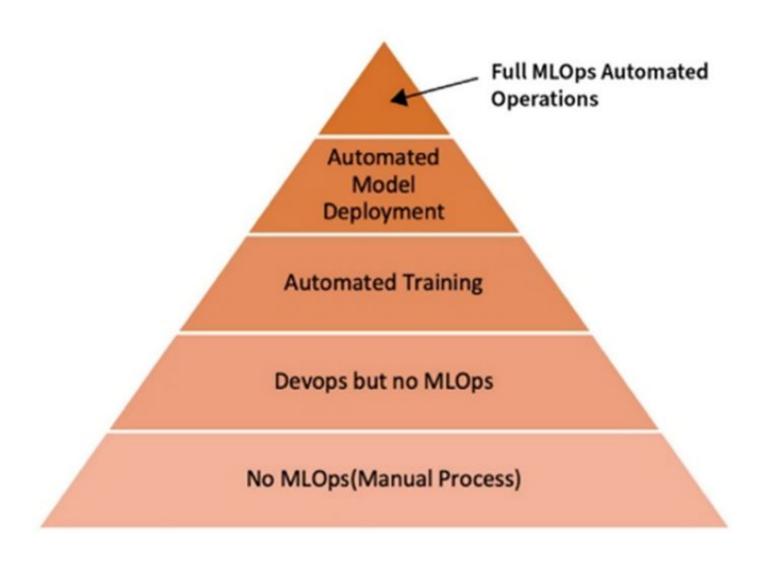






Automatizing the solution: Towards MLOps















How to achieve MLOps: CRISP-ML

Need of a framework: CRISP-ML

- Cross Industry Standard Process
 (CRISP) for Machine Learning
- 2. Proposed by Studer et al. in a paper in 2021(1)
- It is a methodology specifically designed for guiding machine learning projects.
- 4. The process consists of six phases: Business and Data Understanding, Data Preparation, Modeling, Evaluation, Deployment and Monitoring and Maintenance.

ml4devs.com/mlops-lifecycle **MLOps: Eternal Knot** Machine Learning Lifecycle COLLECT FORMULATE Data Dev © 2022 Satish Chandra Gupta CC BY-NC-ND 4.0 International License twitter.com/scgupta





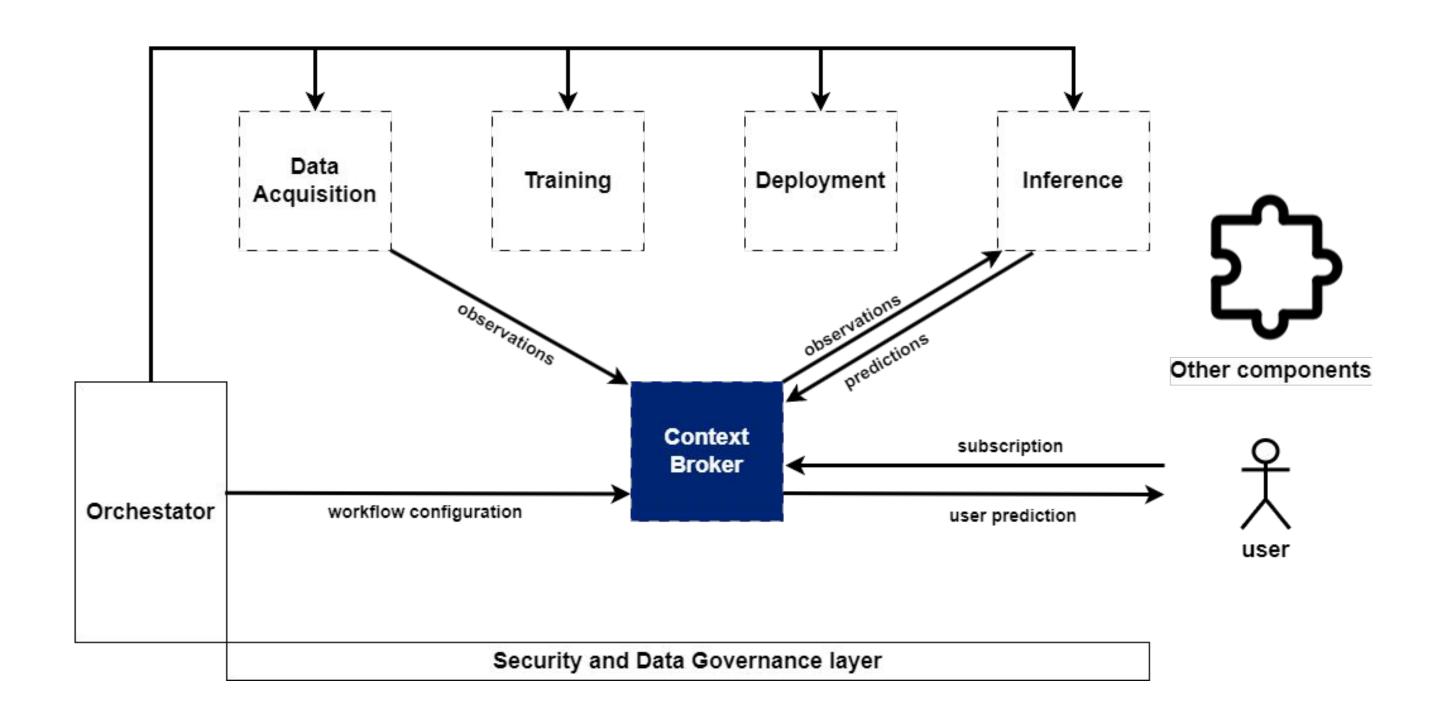








FIWARE approach to MLOps



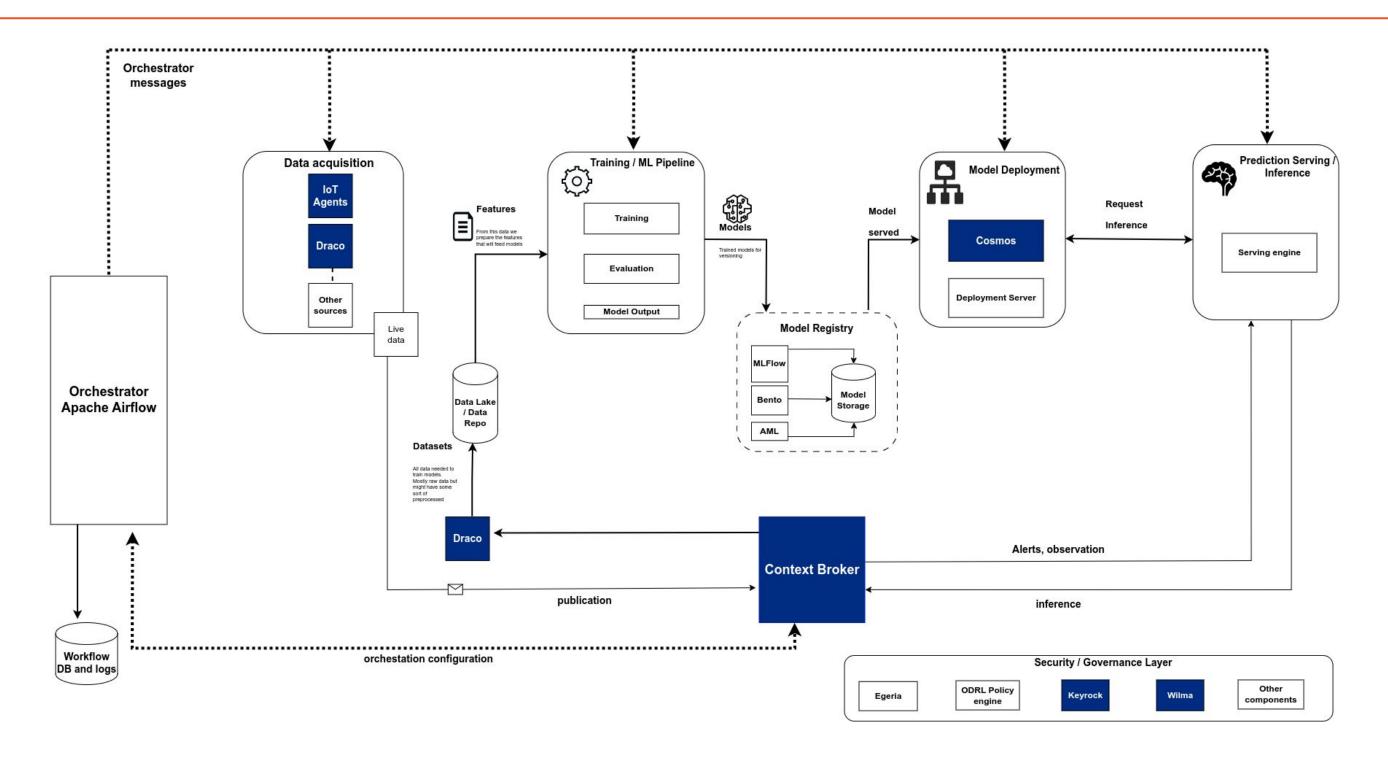






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Entities:

MLModel: describes the model, how was trained, dataset used, algorithm, accuracy, etc.

MLProcessing: It links 3 elements:

- 1. The ML model to be used for the processing
- 2. The data that the ML model will subscribe to
- 3. The entity where the ML model will store the result(s) of the processing

Subscriptions: SubscriptionQuery:

- Inform when there is an attribute update (i.e. when a new inference should be done)
- Inform when there is a prediction update



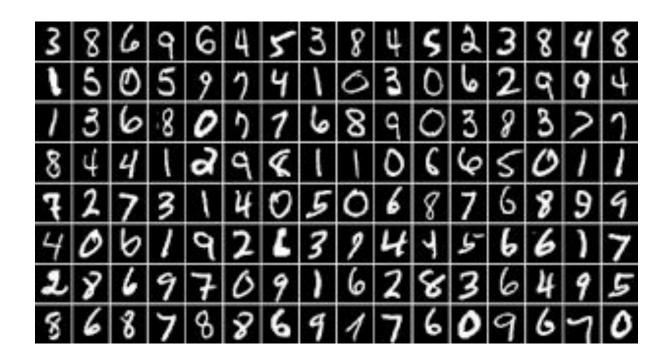




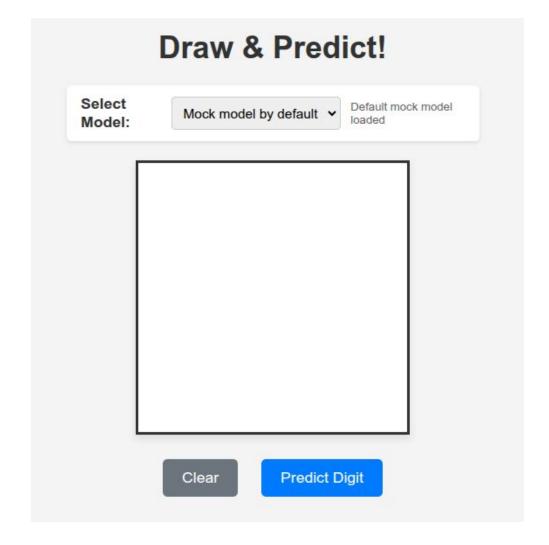




Digit recognition example



http://localhost:5001/predict













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MLModel Entity

```
{
   "id":"urn:ngsi-ld:MLModel:01",
   "type":"MLModel",
   "name": "Handwriting digit predictor",
   "description": "Given an image path, predicts the
digit represented...",
   "dockerImage": "Docker image containing the model",
   "algorithm": "CNN",
   "version": 12,
   "inputAttributes": ["imgPath", "size", "color"],
   "outputAttributes": ["predictionClass", "confidence"]
}
```

 It describes the model, how it works, what the input and output are







MLProcessing Entity

```
{
   "id":"urn:ngsi-ld:MLProcessing:01",
   "type":"MLProcessing",
   "refMLModel": "urn:ngsi-ld:MLModel:01",
   "refSubscriptionQuery: "urn:ngsi-ld:SubscriptionQuery:01",
   "@context": [
   "https://raw.githubusercontent.com/smart-data-models/dataModel
   .MachineLearning/master/context.jsonld"
   }
}
```

- It relates three key elements:
 - The MLModel used to perform inference (who)
 - The data that el MLModel subscribes to process (what)
 - The entity where the MLModel will store the result(s) (where)

The advantage of using a subscription is that we can use something flexible to express queries like "I want the ML model to process certain attributes <a1, a2, ..., an> of an entity <T1> that fulfill certain requirements <r1, r2, ..., rk> (e.g. created by certain user, located in specific area, etc.)







SubscriptionQuery - Inform about

```
"id": "urn:ngsi-ld:Subscription:01",
"type": "SubscriptionQuery",
"entities": [
    "type": "<mark>HandwritingDigit</mark>"
'notification": {
  "attributes": [
    "imagePath",
    "imageMetadata"...
  "format": "keyValues",
  "endpoint": {
    "uri": "http://my-domain-name",
    "accept": "application/json"
```

```
"q": "language==urn:ngsi-ld:lang:Spanish",
    "watchedAttributes": [
        "predictedDigit",
        "confidence"
      ],
      "@context": [

"https://raw.githubusercontent.com/smart-data-models/dataModel.MachineLearning/master/context.jsonld"
      ]
}
```









A typical ad-hoc ML pipeline

Data Scientist

- Get and clean up data
- Prepare and train a ML model

IT person (deployment service)

Package and deploy the ML model

End user / service

- Discover the available ML models (with respect to privacy)
- Ask to use one or more of them (and optionally pay for it)
- Get real time data (predictions, outliers,...) from a ML model

Machine Learning Workflow

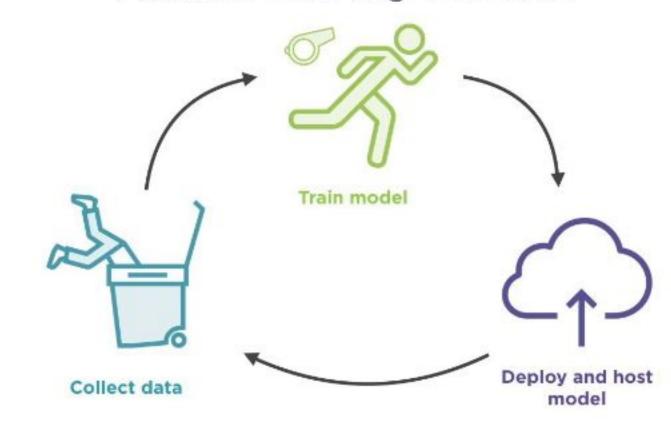


Image taken from here





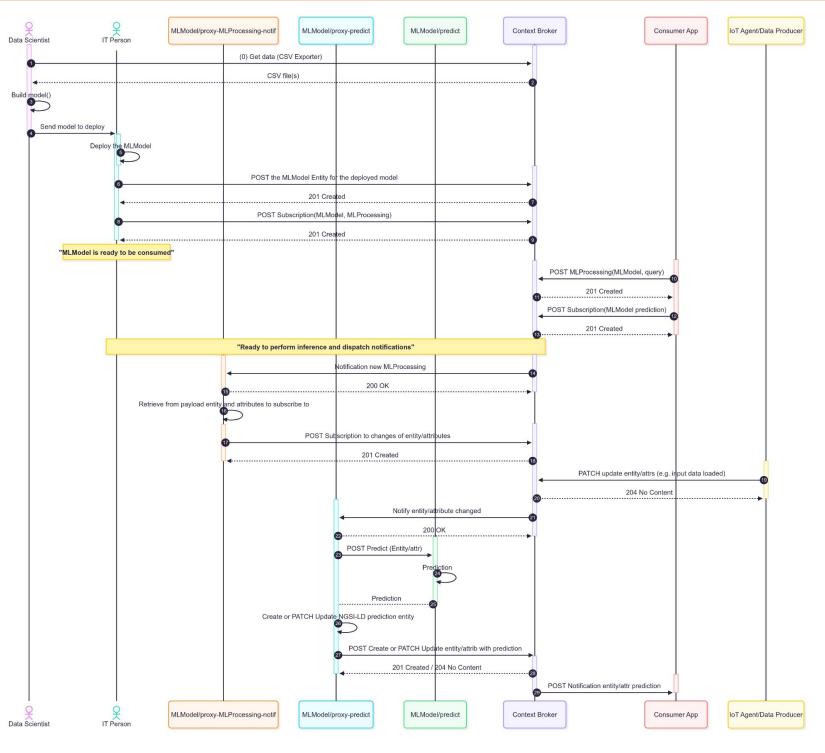








Integration in FIWARE



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Summary

- Described the ML and MLOps process and how differ from traditional SW dev.
- MLOps concepts description and how can be implemented using CRISP-ML methodology
- Proposed a reference architecture for integration of MLOps workflow in FIWARE solutions
- Described Smart Data Models for the characterization of ML deploying and inference process
- Set the basis of a possible workflow for performing MLaaS in FIWARE solutions (SDM and subscriptions)











Resources

- Repository with code and slides:
 - https://github.com/dncampo/fgs2025-digit-recognizer-ml
- MLOps in FIWARE example:
 - Document architecture <u>here</u>.
 - o FGS2024 example and slides: https://github.com/dncampo/FGS2024_ML-training/tree/main
 - Parking example using MLOps concepts: https://github.com/ging/fiware-mlops-parking
- MLOps maturity model from Microsoft, <u>here</u>
- <u>CRISP-ML</u> methodology





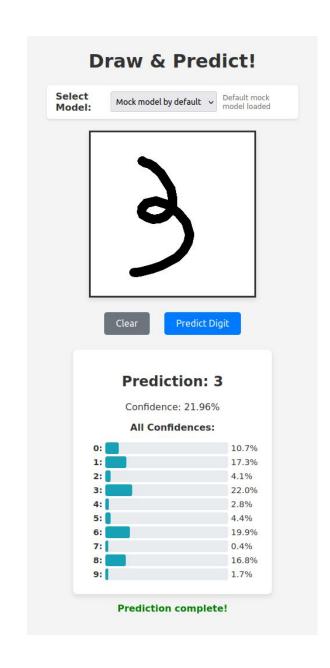






Follow-up / exercises

- Please clone the repo https://github.com/dncampo/fgs2025-digit-recognizer-ml
- Run the app for prediction and send (HandwritenDigit) images, what happen?
 Could you check to against the CB if any entities are there? What is the entity type of these images?
- You will find a trained model...why aren't they used in the app under the /predict endpoint?
- How would you solve that? Implement the expected piece of code / entities to solve the issue
- Now you see a ML model, what do you need to produce some notification in a specific endpoint? Implement the necessary entity to send to the CB to receive the notification after a prediction

















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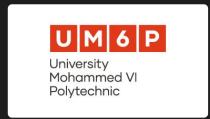








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