





The 2021 International Conference on Artificial Intelligence and Big Data in Digital Era (ICABDE)

# Towards meta-learning based data analytics to better assist the domain experts in industry 4.0

Moncef Garouani<sup>1,2,3</sup>, Adeel Ahmad<sup>1</sup>, Mourad Bouneffa<sup>1</sup>, Mohamed Hamlich<sup>2</sup> Gregory Bourguin<sup>1</sup>, and Arnaud Lewandowski<sup>1</sup>

<sup>1</sup> LISIC Laboratory, ULCO University, Calais, France <sup>2</sup> ISSIEE Laboratory, ENSAM, University of Hassan II, Casablanca, Morocco <sup>3</sup> Study and Research Center for Engineering and Management, HESTIM, Casablanca, Morocco



# Context (1/2)

![](_page_2_Figure_1.jpeg)

- Automation and data exchange in manufacturing industry
- Smart factory
- Data analytics approaches in classical manufacturing factories
- Recent development trends vs Industrial Data Analytics (IDA)

# Context (2/2)

![](_page_3_Figure_1.jpeg)

- Industrial data analytics in Industry 4.0
- Industry 4.0 vs Machine Learning
  - Industry Internet of Things (IIoT)
  - Cloud Computing and Edge data computation
  - production systems in smart factories
  - Interconnected continuous data exchange
  - Data lakes
  - Industrial Big Data Analysis

## Motivation

#### **Industrial Data Analytics**

- Industrial Data Analytics (IDA) provides assistance to engineers in Industrial decision-making.
- These systems provide recommendations in finding the right diagnosis or the optimal decision.

![](_page_4_Picture_4.jpeg)

#### **Industrial Data Science**

- Challenges in building ML predictive models with big manufacturing data
  - Efficiently selecting ML algorithms
  - Efficiently configuring related hyperparameters

- Autonomous industrial smart applications
- Rapid collaboration among industrial actors and data scientists
- Decision-support systems
- Explainability of intelligent predictions

### Introduction

#### **Industry 4.0 vs Machine learning**

Predictive analytics

 $\geq$ 

- Data-driven decision making
- Interpretation of data patterns
- Make value from massive industrial data

![](_page_5_Picture_6.jpeg)

**IDA applications** 

- Process level
- Machine level
- Shop floor level
- Supply chain level

# Introduction

Human tuning process configuration

![](_page_6_Figure_2.jpeg)

# Meta learning (1/2)

□ *Learn to learn* using previous knowledge from related tasks.

□ Algorithm Learning (selection)

- Select learning algorithms according to the characteristics of the instance.
- □ Hyper-parameter Optimization
  - Select hyper-parameters for learning algorithms.
- "Algorithms show similar performance for the same configuration for similar problems"

## Meta learning (2/2)

![](_page_8_Figure_1.jpeg)

# I- Learning phase

![](_page_9_Figure_1.jpeg)

Figure 1: The workflow of the learning phase

![](_page_9_Picture_3.jpeg)

### **II- Recommendation phase**

![](_page_10_Figure_1.jpeg)

Figure 2: The workflow of the recommendation phase

#### Framework

![](_page_11_Figure_1.jpeg)

Figure 3: The workflow of the proposed framework

![](_page_11_Picture_4.jpeg)

**Evaluation Benchmark** 

Dataset	Num. of Classes	Num. of Instances	Task
(Mazumder et al.,2021 )	4	959	Failure risk analysis of pipeline networks
(Benkedjouh et al., 2015)	2	61000	RUL prediction
(Saravanamurugan et al., 2017)	3	2000	Chatter prediction
(Costa and Nascimento, 2016)	2	60000	APS system failure prediction
(Baldi et al., 2014)	2	98050	high-energy physics data analyses
(Tian et al., 2015)	7	1941	Faults detection

Table 1 : List (sample) of Datasets used in the evaluation

• Mazumder, Ram K., Abdullahi M. Salman, and Yue Li. "Failure risk analysis of pipelines using data-driven machine learning algorithms." SF 89 (2021): 102047.

• Benkedjouh, Tarak, et al. "Health assessment and life prediction of cutting tools based on support vector regression." JIM, 26.2 (2015): 213-223.

• Saravanamurugan, S., et al. "Chatter prediction in boring process using machine learning technique." IJMR, 12.4 (2017): 405-422.

• Costa, Camila Ferreira, and Mario A. Nascimento. "Ida 2016 industrial challenge: Using machine learning for predicting failures." IDA. Springer, Cham, 2016

• Baldi, Pierre, Peter Sadowski, and Daniel Whiteson. "Searching for exotic particles in high-energy physics with deep learning." NC, 5.1 (2014): 1-9.

• Tian, Yang, Mengyu Fu, and Fang Wu. "Steel plates fault diagnosis on the basis of support vector machines." Neurocomputing 151 (2015): 296-303.

Evaluation strategy

![](_page_13_Figure_2.jpeg)

<---->

Evaluation results

Dataset	Recommended config. result	Paper result	Pipeline with default config.
(Mazumder et al.,2021)	93.74	85	80.24
(Benkedjouh et al., 2015)	99.41	98.95	93.88
(Saravanamurugan et al., 2017)	97.06	95	86.12
(Costa and Nascimento, 2016)	99.10	92.56	92.34
(Baldi et al., 2014)	85.59	88	69.45
(Tian et al., 2015)	99.54	80.74	76.23

Table 2: Performances of the proposed framework

Evaluation results

![](_page_15_Figure_2.jpeg)

(industrial researchers) configurations.

#### Conclusion

![](_page_16_Figure_1.jpeg)

#### Perspectives $\wp$

The next planned steps include:

1. Further validation of the proposed framework in other real world applications with a larger and more diverse problems .

![](_page_17_Picture_3.jpeg)

2. Add support for further data formats and ML algorithms

![](_page_17_Picture_6.jpeg)

# THANK YOU FOR YOUR ATTENTION

ဗိုဂ္ဂီဗို

![](_page_19_Picture_0.jpeg)

![](_page_19_Picture_1.jpeg)

![](_page_19_Picture_2.jpeg)

# Towards meta-learning based data analytics to better assist the domain experts in industry 4.0

![](_page_19_Picture_4.jpeg)