0.98

OPTIMA Optimizing Manufacturing Processes through Artificial Intelligence and Virtualization.

Industry 4.0 – Digital Transformation in Industry

Theodosis Theodosiou Assistant Professor



PTIMA

May 11, 2022 • Gaiopolis Conference Center • Larissa, Greece



> Optimizing Manufacturing Processes through Artificial Intelligence and Virtualization.



Grant Number	958264
H2020 Call	H2020-NMBP-TR-IND-2020-singlestage
Duration	1 January 2021 – 31 Dec 2023 (36 Months)
EU contribution	€ 8.359.087,50
Countries	Greece, Cyprus, Germany, Finland, Italy, Spain, Ireland, United Kingdom

Principal Investigator for UTH: Prof. E. Papageorgiou

✓ 16 Partners✓ 8 Countries



Motivation and Objectives

Motivation:

> Modern industry has a saying

«Cheap, Reliable, Fast Choose two out of threee.».

 There is always a compromise between production time, quality and resources need in the production process.





Motivation and Objectives

Objectives:

Reduce compromise by:

- Installation of smart sensors that implement
 Artificial Intelligence methods for monitoring the production process and quality control.
- Digital representation of the production in a virtual environment (digital twins) aiming the simulation, the optimization and the planning of production.





OPTIMAI

Basic Idea

Steps:

- Installation of Smart Sensors along the production line.
- > **Real-time** production monitoring.
- > Use of Artificial Intelligence το detect and diagnose faults at early stages.
- > Use of **Digital Twins** and simulations.
- Use of Augmented Reality to deliver messages and control production.



Architecture

OPTIMAI is a modular platform for management and planning the production.

Its functionality relies on the collaboration of **multiple modules**:

- Quality Control
- Artificial Intelligence
- > Digital Twins
- > Production Optimization
- Decision Support
- > etc.





Use cases



Zero-Defect Quality Inspection

Monitoring the production targeting **Zero-defect Manufacturing**.

ØPTIMAI



Production line setup-calibration

Fault detection and prediction. Real-time data analysis using Artificial Intelligence for (re)configuration/calibration of machinery.



Production Planning

Scenario Exploration. Use of simulations and digital twins towards a more efficient production planning.



Pilot sites

OPTIMAI will installed and employed at three Pilot sites:

Microsemi (Ireland): Fabrication of microchips.

> Goal: Detect defective dies, identify cause of defects.

KLEEMANN (Greece): Manufacturing of elevators

> Goal: Detect sub-optimal operation of the Power Hydraulic Unit.

Televes (Spain): Fabrication of antennas

> Goal: Detect defective antennas (damage, assembly).



Artificial Intelligence

for defect detection

@DTIMAI

Defect detection and analysis on wafers (1/2)

Dies are produced in batches on wavers. A special probe identifies defective dies and produces a visualization of the distribution of defects (wafer map).



Statement of the problem:

- (1) Recognize patterns of defects.
- (2) Correlate patterns with production parameters to identify cause of defect.

Suggested solution

Convolutional Neural Networks (CNNs)



Defect detection and analysis on wafers(2/2)

Approach & Implementation:

- > Implementation of numerous architectures.
- > Assessment of architectures and models.
- Selection of optimal solution in terms of

Accuracy, Reliability, Speed.







Defect detection in elevator hydraulic power unit (1/2)

The HPU is assembled and operated. An expert technician monitors a groups of measurements (pressure, velocity, noise) to determine if operation is optimal or not.



Statement of the problem:

Operational parameters should be considered in combination

- > Time-consuming and error-prone process.
- > No defective timeseries available.

Suggested solution:

CNN- and LSTM-based Autoencoders.





Defect detection in elevator hydraulic power unit (2/2)

Approach & Implementation:

- > Assessment of existing methodologies.
- Due to the lack of defective products, the model had to "learn by itself" if an operation is optimal or not.

			Enco	der	Decoder						
$\underline{\bigwedge}$	Conv1D 64	Conv1D 32	Conv1D 16	Conv1D 8	Max-	Transp Conv1D 16	Transp Conv1D 32	Transp Conv1D 64	Transp Conv1D 64	Conv1D 3	$\underline{\Lambda}$
	ReLU	ReLU	ReLU	ReLU	pooling Size 2	ReLU	Rel U	Rel U	ReLU	Linear	
- Nacional Anna - Anna	Size 15	Size 9		Size 3	Sride 2	Size 3	Size 7	Size 9	Size 15	Size 1	- Next
	Stride 3	Stride 2	Stride 2	Stride 1	8×8	Stride 2	Stride 2	Stride 2	Stride 3	Stride 1	
			17×16	17×8		16×16	32×32	2010- padding (2,1)		201x3	
X _{201×3}	67×64	34×32						6/×64	201×64	201/0	$\widehat{X}_{201 \times 3}$
					L2 Lo	ss					

CNN AE

Detector	Accuracy
WT-CNN classifier	96%
LSTM-based AE	91%
CNN-based AE	94%



Defect detection in antennas

Antennas are produced in an automated production line.



Statement of the problem:

Defects during assembly (wrong parts). Defects due contacts (misalignment, cracks).

Suggested solution: Use of AI – Task in progress.



Conclusions

Conclusions

Artificial Intelligence methods:

- can automate the detection and diagnosis of defective products.
- > can "understand" defective products without prior training.
- > do not replace human intelligence but simplifies operations by automated repeating and time-consuming processes.





OPTIME

Information - Contact

Theodosis Theodosiou Assistant Professor dozius@uth.gr Elpiniki Papageorgiou Professor elpinikipapagorgiou@uth.gr



This project has received funding from the European Union's Horizon 2020 research and innovation programme under Grant Agreement No. 958264 The material presented and views expressed here are the responsibility of the author(s) only. The EU Commission takes no responsibility for any use made of the information set out.