ForeSee Cluster Webinar Series

Predictive Maintenance: European success stories



European Cluster for Sustainable Predictive Maintenance Solutions in the Factory of the Future



ForeSee is a cluster of six projects, which have received funding from the European Union's Horizon 2020 research and innovation programme under the FoF-09-2017 call.















European Cluster for Sustainable

Predictive Maintenance Solutions

in the Factory of the Future

Webinar #2

Reliability and Tools, for Predictive Maintenance Applications Professor Dr. Basim Al Najjar Linnaeus University, Sweden Thanasis Naskos, Atlantis Engineering, Greece



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PdM with respect to:

- Accuracy
- Applicability
- Data availability
- Cost-effectiveness



PreCoM: Addressed problem and Purposes

- How to reduce the losses in production and maintenance that other systems did not succeed to achieve?
- Development of Predictive Cognitive Maintenance Decision Support System (PreCoM-system) to improve availability and maintainability, reduce accidents and losses in energy and material cost-effectively.





Predictive Cognitive Maintenance Decision Support System





PreCoM- Smart PdM platform and philosophy



- To describe reality, i.e. machine condition, we better apply a combination of:
 - Deterministic (what has been happened and ongoing), and
 - Probabilistic (what may be happened conditionally) approaches.
- Well identified, high quality and better coverage data to provide better understanding of the real condition of a Machine – Component – Deterioration process – More accurate prediction and Maintenance decision.



Tools, methods and technologies developed and utilized



- Data driven, big data and statistical models,
- Modules for automatic:
 - Data acquisition,
 - Detecting damages,
 - Localizing and assessment of damages,
 - Condition diagnosis,
 - Predicting damage development,
 - Recommending about; what, where, why and how to do,
 - Detecting deviations in the health of sensors,
 - Assessment of time to maintenance and optimizing maintenance for the whole machine,
 - Production scheduling and optimizing maintenance w r t a production schedule.





- AR and PLIV,
- Tools to digitalize manually gathered data,
- Module to provide statistics to the machine manufacturer supporting modification of new generations,
- Module to follow up and assess PdM technical and economic impact on production process,
- Common data and knowledge base,
- Common API and Cloud for communication



Whom utilize them



- 1. Maintenance personnel for maintaining machine condition effectively, and
- 2. Production personnel for effective and more profitable production.





Application: Case studies

- PreCoM-system is applied on four machines, in three different companies in two EU-countries (Spain and Slovakia).
- The machine condition, production, etc. have been monitored 24/7 during the demonstration period.
- Data has been collected continuously for some modules and periodically for others.
- Analysis, results and recommendations are provided directly after every measuring opportunity.
- Follow up and treatment of problems, challenges, bugs and deviations were done continuously to secure smooth performance.













Challenges and Solutions



- **Communications**: 17 partners of different background, work culture, languages and specialities.
- Development of a common understanding of the problems addressed and purposes to meet project expectations.
- Identification and gathering of relevant, wide coverage and high quality data.
- Development of common cloud/data- and knowledge base and API to secure communication/integration of PreCoM Cloud with all hard- and software modules

We succeeded to **overcome the challenges** through:

- Closer, reliable and committed collaboration of PreCoMpartners.
- Solving problems jointly in a friendly and constructive climate.
- Partners commitments.



Primary results



Intermediate assessment with demonstration companies concerning:

- Accessibility: Easy
- Usability: Good and Reliable
- Impact of the PreCoM system: Problems with components were early detected by PreCoM in two companies (it would not had been otherwise) and led to avoid production stoppages and saving costs for about 20,000-40,000 € per company.

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European Cluster for Sustainable **Predictive Maintenance Solutions** in the Factory of the Future









• Smart Maintenance Process



Data Collection

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Data Pre-processing



Data Processing



Data Analysis



Actions Orchestration



User Feedback



• Sample of Available Tools





• 1st step of Data Processing





• Supervised or Unsupervised ML?

Unsupervised learning is a type of machine learning that looks for previously undetected patterns in a data set with no pre-existing labels and with a minimum of human supervision.



Supervised learning is the machine learning task of learning a function that maps an input to an output based on example input-output pairs. • Proactive or Reactive Maintenance?

Proactive maintenance is a preventive maintenance strategy. The purpose of proactive maintenance is to see machine failures as something that can be anticipated and eliminated before they develop.



Reactive maintenance *refers to repairs that are done when equipment has already broken down, in order to restore the equipment to its normal operating condition.*

Should we choose?



































• Z-BRE4K project



a novel predictive maintenance platform to eliminate unexpected-breakdowns and extend the life of production systems.

This project has received funding from the Horizon 2020 Framework Programme of the European Union under grant agreement n° 768869





Gestamp

The GESTAMP-Autotech demonstrator will be linked to the demonstration of a Lighthouse manufacturing process FRAMETOP for the multi stage zero defect manufacturing of next generation automotive chassis.

Philips

As a complex demonstrator PHILIPS contributes a production line with cold forming tooling. On this tooling PHILIPS would like to move from preventive maintenance to predictive maintenance.

Sacmi-CDS

The SACMI-CDS use case will be related to the packaging industry. SACMI is planning to introduce a predictive maintenance service for the end user involved (CDS) within Compression Moulding machines.

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