CISTER - Research Center in Real-Time & Embedded Computing Systems

Framework for Proactive Maintenance in the Real World

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MPMM 2018 1



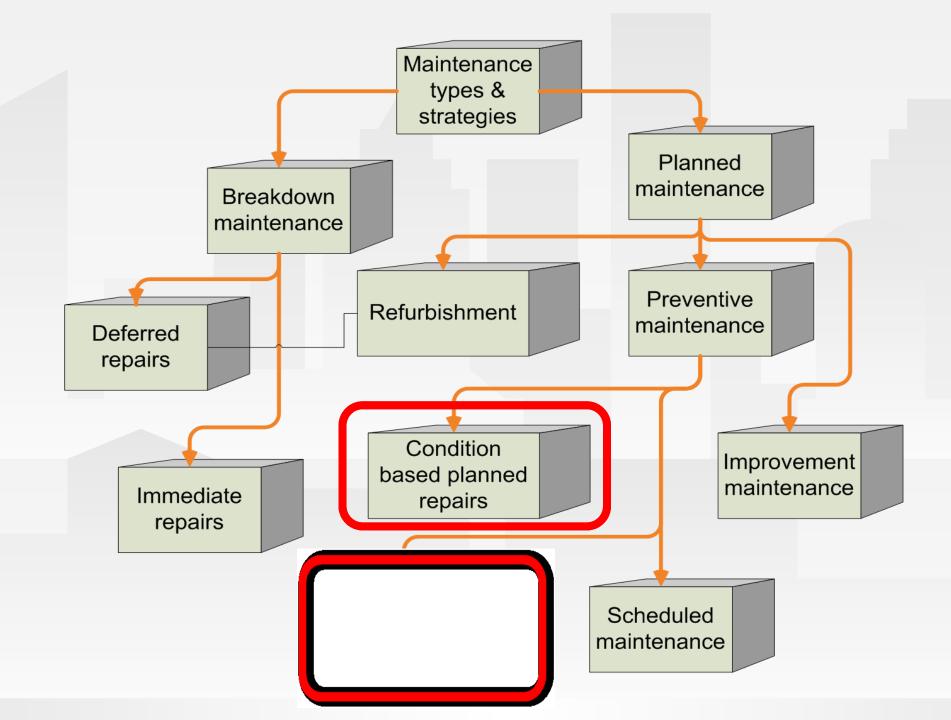
- Modern maintenance
 - Goals
 - Supporting technologies: IoT, CPS, cloud computing
- A platform for maintenance
- Applications in the real world
- Future of maintenance



Different Kinds of Maintenance

- Corrective Maintenance (CM)
 - also called Run-to-failure reactive maintenance, and fire-fight approach
 - Advantage: minimizing manpower to keep things running
 - Disadvantage: unpredictable production capacity, high overall maintenance costs
- Preventive Maintenance (PrM)
 - periodic maintenance execution that can range from equipment lubrication to replacement
 - based on specific periods of time, amount of machine usage (number of working hours) and/or mean time to failure (MTTF) statistics.
 - Advantage: it improves equipment lifetime, it reduces malfunction probability
 - Disadvantage: may occur prematurely, or too late (failures can occur)
- Predictive Maintenance (PdM), or Condition-Based Maintenance
 - relies on physical measurements of the equipment condition (e.g.: temperature, vibration, noise, lubrication, corrosion)
 - maintenance happens in a need-based when a certain threshold is overcome.
 - It is a tool to improve corrective and predictive maintenance
- Proactive Maintenance (PM)
 - builds on PdM, focuses on the root causes of the problems, by modeling the machines and their environments, and comparing with data from the machines
 - depends on the availability of an efficient and effective monitoring infrastructure







In CBM the maintenance engineer needs answers to questions like:

Which machines need maintenance?

What are the necessary actions?

Where is it located? How can I identify it? When should I do this action? What spare parts do I need?

How do I carry out the work? How should I report the work?

It's all about information







Technologies to the rescue

- Development of processing power & cheap sensors
- High interest to IoT, Cyber-Physical Systems
 - Manufacturing companies want to widen their business to maintenance i.e. provide services for their products
 - Companies want to improve their productivity with higher Overall Equipment Effectiveness (OEE)
- Big data, 3 V's (Volume, Velocity, Variety)
 - Need for automatic diagnosis, AI
 - Knowledge Discovery in Databases (KDD)



Internet of Things (IoT)

- Part of the "Embedded" family (CPS, Wireless Sensor Networks, Machine to Machine communication, Body Area Networks, etc.)
- IoT considers all embedded systems as connected to the Internet
 - Allows for ubiquitous access to the embedded systems
 - Makes use of mature protocols and well accepted software libraries, thus:
 - IoT applications are faster to implement and easier to maintain
 - lower time to market and cheaper maintenance



Cyber-Physical System (CPS)

- Term introduced in 2006
 - Has currently many definitions, such as:
 - Technologies that allow to manage interconnected systems between physical assets and computational capabilities
 - Systems of collaborating computational entities that are connected to the surrounding physical assets providing and using services to/from the internet
 - A system consisting of <u>computational</u>, <u>communication</u> and <u>control</u> <u>components</u> combined with physical processes
 - A bridge between the physical and cyber worlds

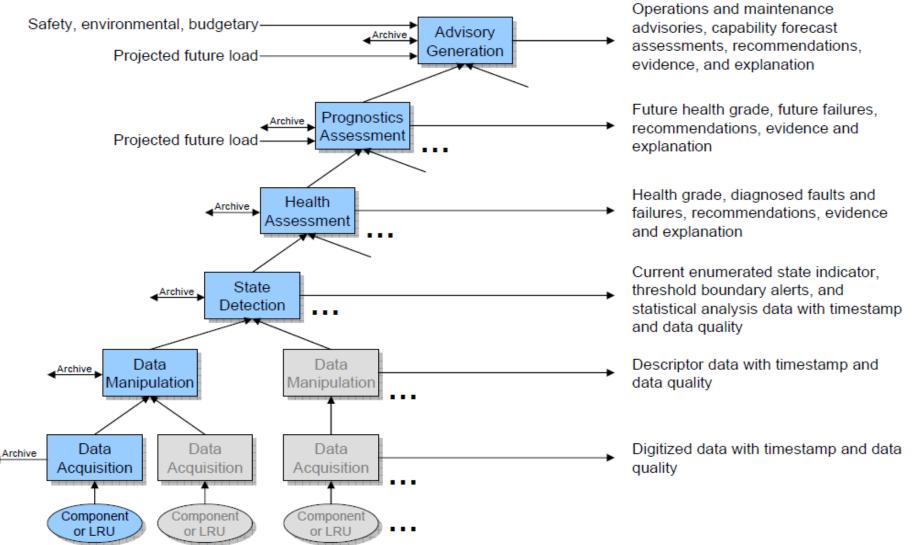
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- High-performance computing
- Well-structured architecture
 - Allows for
 - easy deployment of new applications,
 - elasticity,
 - maintanability / general management
- Well-accepted technology
- Well documented development platforms



ISO 13374: Condition Monitoring and Diagnostics of Machines

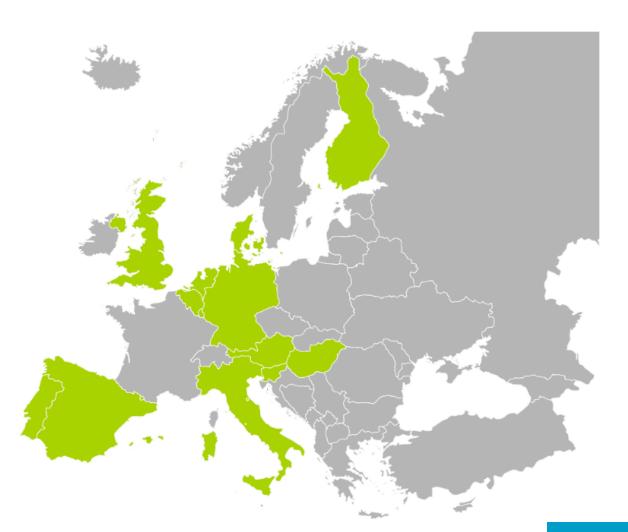


The MANTIS project

EU Joint Undertaking

Funded by ECSEL program and national agencies

47 partners from 12 countries



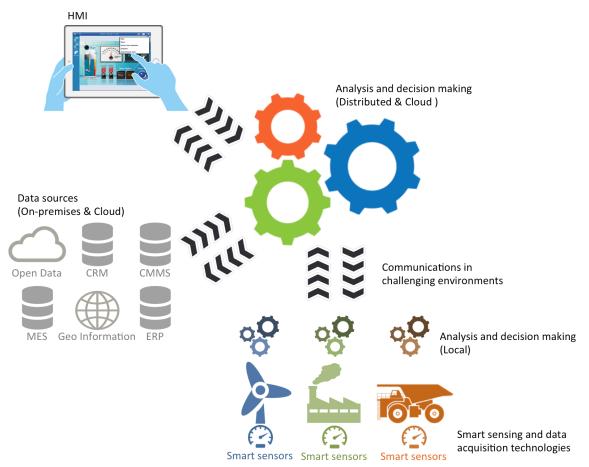
Objectives of the MANTIS

- The use of data structures that <u>enable the collection</u> <u>of maintenance information</u> (events, root cause analysis, fault prediction and remaining useful life results ...)
- The use of <u>data structures that enable large volume</u> of <u>data to be processed</u> in real time or in batch processes.
- <u>Integration</u> of complex and heterogeneous largescale distributed systems from different application domain.
- The design of CPS-populated systems to enable <u>collaborative proactive maintenance strategies</u>.

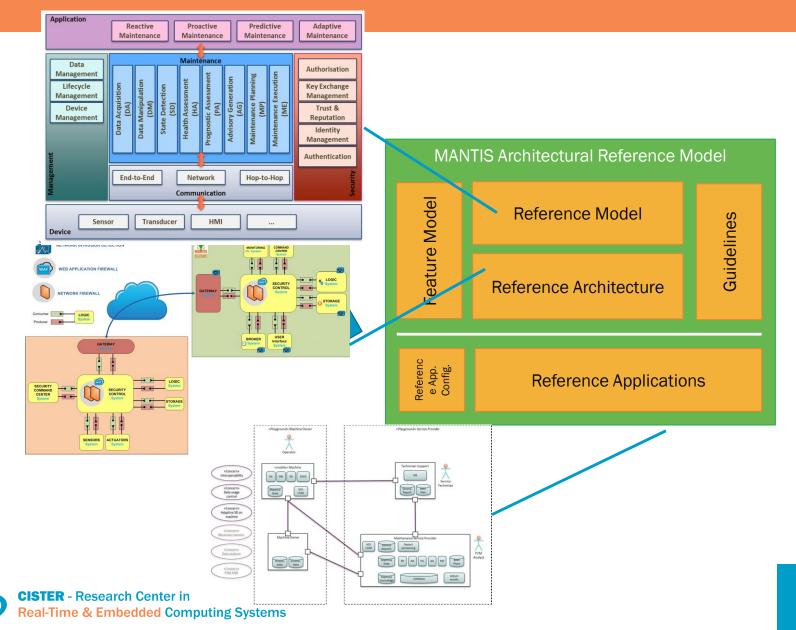


The MANTIS platform

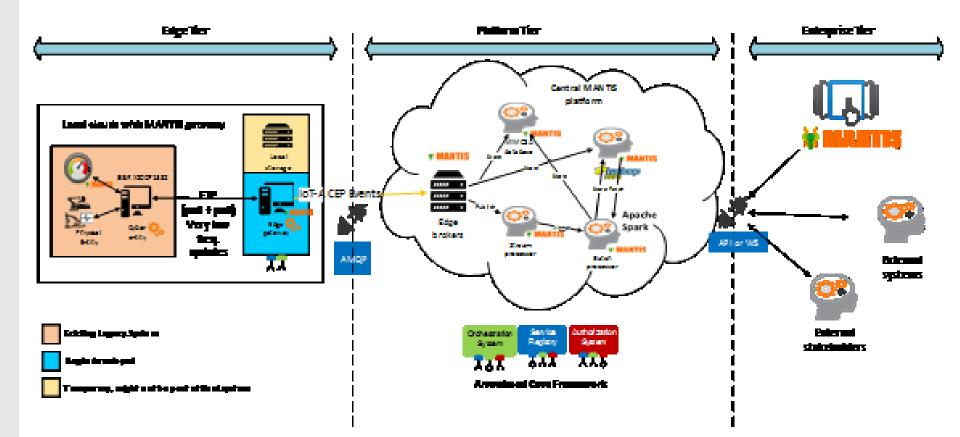
- Developed a framework for proactive maintenance, based on:
 - real-time sensor data acquisition
 - high-performance communication middleware, connecting local networks in the factories and the cloud through challenging environments
 - data analysis and decision making, both local (in the factory) and on the cloud
 - advanced HMI



Reference Model of the MANTIS



Reference Architecture of the MANTIS





Edge Tier elements

- Edge computing:
 - optimizing cloud computing systems by performing data processing at the edge of the network, near the source of the data
- It can be used to
 - reduce the data sent to the platform
 - secures the machines inside the factory from the outside
 - Supports additional protocols to connect to other devices
 - Converts between different data formats
 - enable on-site maintenance operations with low latency,
 - allow for maintenance operations when connection to the cloud is absent / challenged.

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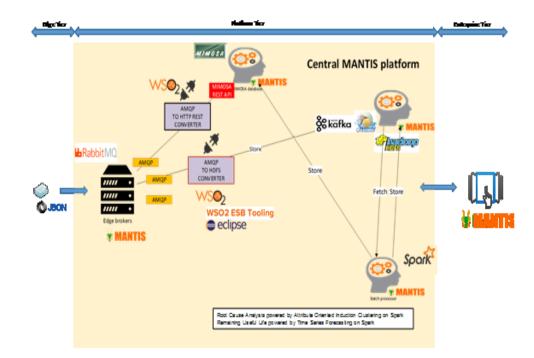
CBM at Low level (Edge)

- The edge tier provides for
 - Local data storage
 - Limited analytics
 - Local HMI
 - Lifecycle management support.
- CPS are involved and are executing their general production operations.



Platform Tier elements

- Data Access and Ingestion through the Edge Broker
 - Publish/subscribe servers
 - Translator/Converters
- Data Storage systems
 - Mimosa
 - Distributed File Systems
- Batch Processor
- Stream Processor

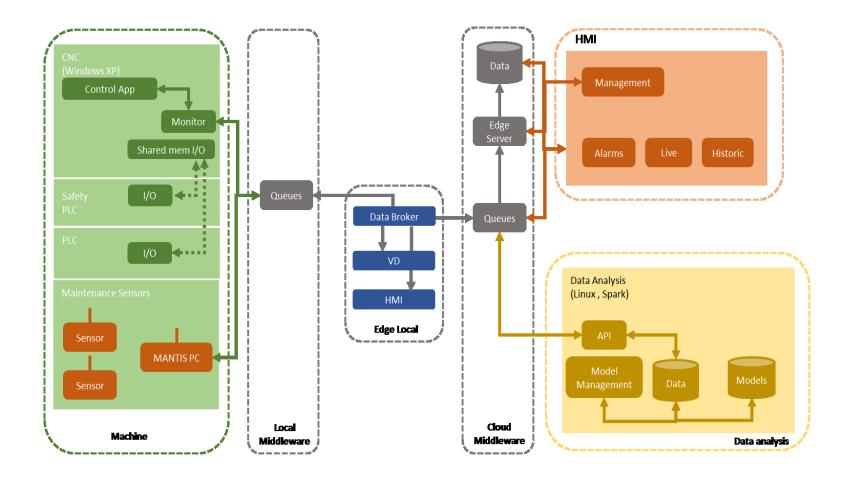


CBM at High level (Platform)

- Remaining Useful Life (RUL) of components: continuous tracking of telemetry (usage) data and estimating how much time the given device or component has left before needs to be replaced
- Fault Prediction (FP): the system shall predict based on diagnostic data an inbound failure mode (different to wear-out to be detected by RUL),
- Root Cause Analysis (RCA): when an unpredicted, complex failure occurs, the system shall deduct the actual module (cause) that caused the break.



Concrete Architecture of the MANTIS





Components: Machine

- Machine
 - PLC: that controls the machine no direct access
 - CNC controller: HMI to configure and operate the machine
 - Already capable of producing basic diagnostic functions
 - Sensors, both already on the machine, or added



Components: local middleware

- Local middleware
 - Connects all factory machines with the Edge Local
 - Can be implemented in different ways.
 Examples:
 - REST and AMQP
 - OPC-UA over AMQP
 - Trendy tech: OPC-UA enables the connection of the machine with other IT systems: MES, SCADA, ERP, etc.

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Components: Edge Local

- Edge Local
 - Secures the machines inside the factory from the outside
 - Supports additional protocols to connect to other devices
 - Connects to the cloud
 - Converts between different data formats
 - Focus on Data: data storage, visualization with local HMI, data filtering and aggregation

Components: Cloud Middleware

- Current trend: storing all data using a MIMOSA-like design
 - High potential for interoperability
- Intermediates with the Data Analysis and HMI modules
- Makes available a set of services which are used by the HMI to configure the system, used to support system start-up and resume system operation in case of a crash
- Communication should be based on highperformance messaging bus
 - Example: AMQP / RabbitMQ

Components: Data Analysis

- Data Analysis
 - Prediction Models:
 - used for the detection, prognosis or diagnosis of machine failures. The models can be built for one machine family, or can be generic and adapted to different machine families.
 - Data and Prediction Interface
 - outputs predictions (to th HMI) from the models, and provides data to feed and train the models.
 - Intelligent Maintenance Decision Support System (IMDSS)
 - used to manage the models (model generation, selection, training and testing), IMDSS is composed of:
 - Knowledge Base of diagnosis and prediction models
 - Rule based Reasoning Engine which includes all the rules that are necessary to deduce new knowledge - helps the maintenance to diagnose failures

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Issues with Data Analysis development

- Errors are extremely rare, 1 error every 150 days
 - impacts the performance of the models, both in terms of learning and prediction accuracy
- Some machine show a failure rate of once every 500 days, therefore we opted on short term predictions via real-time stream analysis
- Communications infra-structure takes care of fusing data from various sources (controller signals, controller sensors, external wireless sensors, etc.), insuring synchronization and time stamps data.
 - These timestamps are used by the failure detection algorithm to label the signal indicating were the failure occurred and allow visualisation by the HMI,



Components: HMI

- Human Machine Interface
 - Controls system start-up
 - Visualization of data
 - Shows analysis results
- Main issue of data vs information



Focus on Sensors and CPS

- CBM uses intelligent techniques, such as machine learning
- Every technique depends on data on the working condition of the machine, and on the environment
- Data is
 - Collected by means of sensors
 - Preprocessed by means of sensors
 - Used for early decisions in the factory by the sensors
 - Communicated through challenging environment to gateways by the sensors













Categories of Sensors

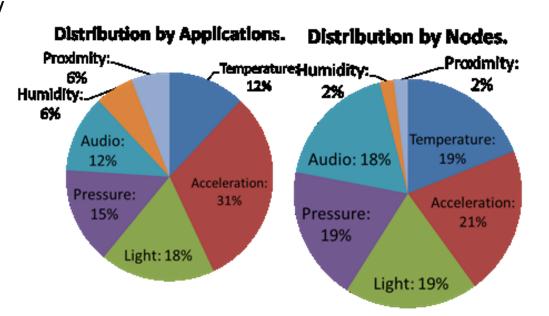
- Sensors can be categorized using different approaches
- A pragmatic approach (in relation to real life use cases) considers the sensors as physical elements, or virtual ones
- With the same pragmatism, physical sensors are either
 - Off-the-shelf (bought)

– Custom (built)

• Virtual sensors are always built (programmed)

Off-the-shelf Sensors

- Measure very common parameters
- Previous works already analysed types of sensors by frequency of usage
- Seven sensor types were identified
- All these sensors are better bought
 - Connected with CPS platforms (Arduino, Raspberry pi, etc), e.g. by I2C



Custom Sensors

- They are specialpurpose
 - E.g.: crack detection, torque easurement, analyse wear of material and retrieve oil status



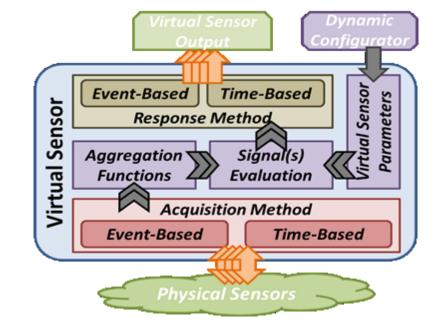


- Not mass produced
- "Buy" would thus be very expensive
- Better to "build"



Virtual Sensors

- They work on data collected by other sensor(s). Usually
 - They are software
 - They "run" in the factory
- Aggregation Functions
 compute new data
 - By combining data from multiple sources
 - By applying process models
 - By considering time series
- Signal(s) Evaluation takes decision / evaluates results

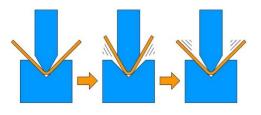


Sheet Metal Bender pilot



Mantis in Portugal: Sheet Metal Bender

- Press Braking: deforming a metal sheet (workpiece) along a given axis by pressing it between clamps (tools).
 - Workpiece from 0.6 to 50 mm thick and lengths from 150 mm to 8 m long
 - In order to have a finished part, a metal sheet will be consecutively bent at several places – e.g. to make a computer box.
- The machine is able to exert a force up to 2200 kN using 2 electric motors of 7.5 kW each, and it is able to bend metal with high precision.
- Compensates for the bending of the ram and structure deformation in order to achieve very high precision.





Implementation

Edge Tier:

- Acquire internal machine sensor and actuator data
 - 30 ms period, 50 variables, for every machine cycle: ≈ 300 MB/day
 - Data are on an Access DB file on the CNC
- Add maintenance specific sensors to the machine:
 - Accelerometers on the blades
 - Oil sensors
- Join all data from a factory on a Edge computer,
 - Preprocessing
 - Local HMI

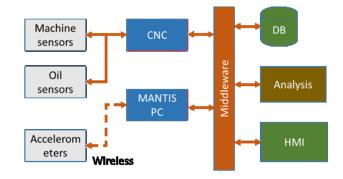
Platform Tier:

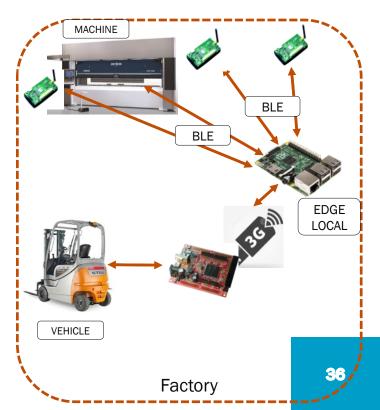
- Transfer all information for analysis to the cloud
 - Communication Middleware based on RabbitMQ
- Analyse data using Machine Learning

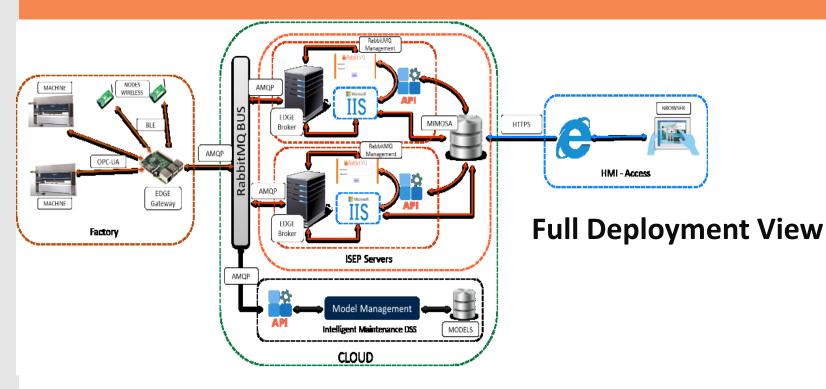
Enterprise Tier:

- HMI to visualise data from the machines
 - Sensor data
 - Maintenance-related data

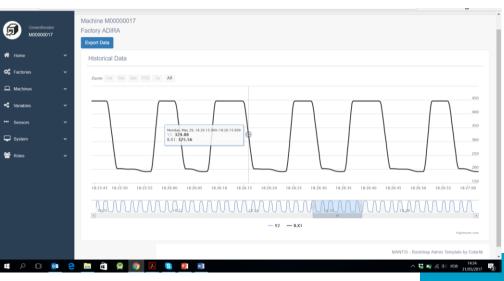
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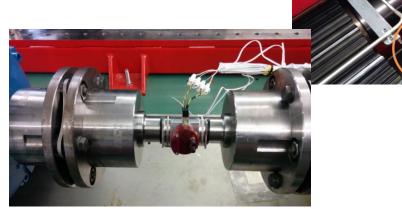
Human Machine Interface



Mantis in Spain 1: Press Machine

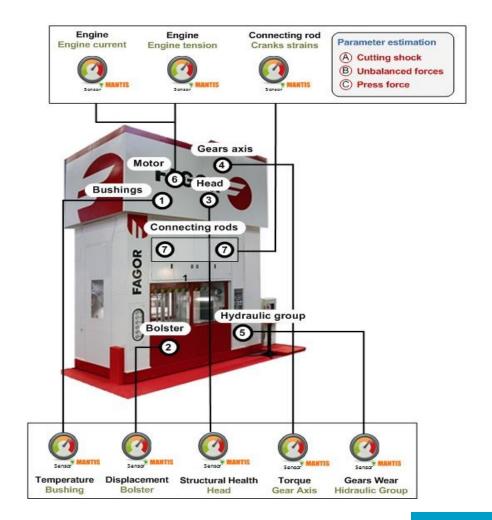
- Cut/shape metal with a die
- Target:
 - 40 million
 strokes, each
 one with a force
 of 2000TN
 - High precision
 - High availability
- Incorporated CPS into most critical components





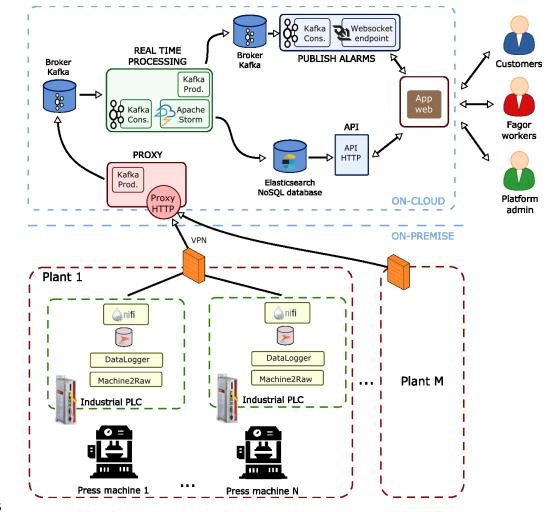
Components requiring CPSs

- Bushings (Temperature and oil condition status)
- Bolster (Relative displacements)
- Head (Structural health)
- Gear axis (Torque)
- Engine (Tension and current)
- Connecting rod (Displacement, forces)



Implementation

- Edge Tier (On-premise):
 - Machine2Raw: extracts the data from the different PLCs
 - Datalogger: stores data in a local database (SQL server)
 - Apache NIFI: processes the raw data stored in the local database. After processing the data, Apache NiFi transfers the data to the cloud through the Proxy
- Platform Tier (On-cloud):
 - Elastic Search: (NoSQL database)
 - Apache Kafka: distributed queue message system to decouple the different applications
 - Proxy: published data to different Kafka topics
 - Real Time Processing: three objectives: Extracting raw data categorised using Kafka topics, persisting data in elastic search, executing data analytics to detect possible alarms
 - API: REST API that connects to a web front end
- Enterprise Tier:
 - Visualizes data using web applications





MANTIS in Spain 2: Pultrusion

- Composite glass fibers, basalt, carbon, natural fibers are impregnated with resin and passed through a heating line.
- Result: highly strengthene composite structures with low weight, elevated mechanical and chemical resistance, and electrical and thermal insulation.
- Material used for example for waterproofing Pajares tunnes (Asturias, Spain), for Valencia Lighthouse (Spain), for the pedestrian bridge in Madrid (Spain)



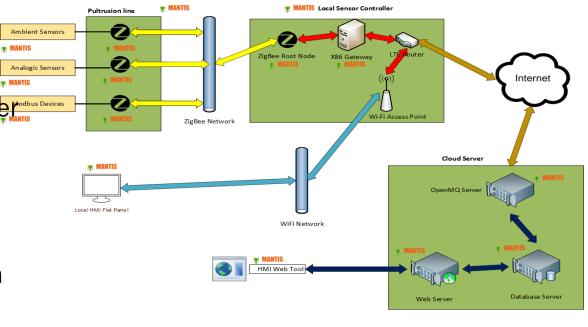
The Pultrusion Machine

- Pultrusion is a continuous process for manufacture of composite materials with constant crosssection.
- Mantis collects data using different sensors in the workshop:
 - Temperature, humidity, air extraction capacity, etc
- and installed in the machine itself:
 - Injection System
 Temperature, Oil tank
 presses system, etc



Mantis implementation

- Different parts / subsystems:
 - Sensors (edge tier)
 - Local Sensors Controlle Controlle Controlle Controlle Controlle Controlle Controlle Controlle Control Con
 - Zigbee[4] Root Node
 - X86 Gateway
 - LTE Router
 - Wi-Fi Access Point
 - Local HMI (edge tier, in the bottom figure)
 - Cloud Servers (platform and enterprise tiers)
 - OpenMQ Server (message-oriented middleware platform, reference implementation for the JMS)
 - Database Server
 - Web Server



😽 MANTIS									Gaccion	
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Peróxidos	26.03 °C	37.76 %	Móvil 1	26.03 °	C 3	37.76 %	Baño 1	26.03 °C	37.76 %	
Mezcla	25.00 °C	41.39 %	Móvil 2	25.00 °	00 °C 41.39 %		Baño 2	25.00 °C	41.39 %	
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MANTIS in the Nederlands: Medical Devices Maintenance

- Medical devices performing non-invasive patient diagnosis
- Medical device must be kept in optimal operational conditions
- Device is life-critical
- Device cannot fail when in use, since
 - not financially possible to have redundancy
 - moving of patients to another hospital might not be possible



- Strategy:
 - complete range of off-theshelf sensor type
 - data are distilled into more advanced information by means of virtual sensors

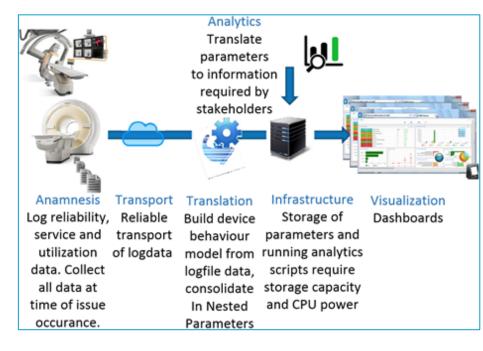
How it is done

Every systems contains many sensors and generates large log files daily

- Anamnesis: optimize (filter) logging
- Transport: make all data available worldwide in the cloud
- Translation: different machines' data to be translated to behavioral models and consolidated in a limited set of relevant parameters
- Infrastructure: the translation requires significant computing power and storage space
- Analytics: the obtained parameters have to be analyzed with respect to the maintenance challenges
- Visualization: the results have to be visualized for end-users

Solution: The e-Alert controller

- Can sample connected sensors (max 8 sensors per interface box)
- Can be connected (daisy-chained) for scalability

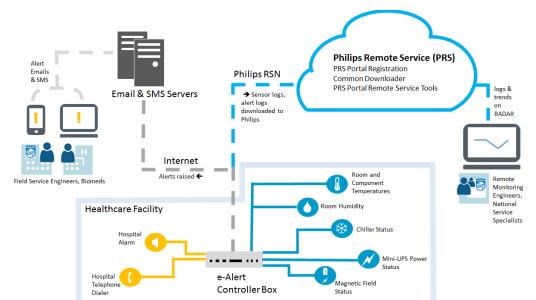




e-Alert controller software

The e-Alert controller provides:

- A web-based user interface
 - to configure sensors (e.g.: thresholds)
 - to see sensor data (online and history)
- Email/SMS send server and Email/SMS receivers
 - to connect to the Philips Remote Service (PRS)
- PRS can:
 - Aggregate and statistically analyze sensor values
 - determine an operational profile, specific to that medical device, to finetune the configured alert thresholds for that specific device



Comparison of the use cases

- Different use cases had different requirements, and satisfied by different kinds of techniques
- All use cases feature real world factories and installation
- Sheet Metal Bender
 - Protects a traditional machine as a whole
 - Machine already having many sensors, but easy to add new sensors
 - Exploited composition of data from off-the-shelf sensors
- Press Machine
 - Protects a machine with close to no sensors
 - Targeted each critical part of the machine
 - Better to "build" custom sensors integrated with the critical parts
- Pultrusion Machine:
 - Huge machine
 - All the shop floor was sensorized
- Medical Device
 - Life-critical machines
 - Deployed in existing networks (hospitals)
 - Important to
 - Filter data, scale the data collection, adapt to different machines
 - Real processing of the data in the cloud

FUTURE OF MAINTENANCE STRATEGY

Provocative questions:

- Is it cybernetic or is it human?
- Is real-time feasible?
- How to determine granularity in space and time?
- Open (customer-in-the-loop) or closed maintainability?
- How to employ advanced communication paradigms (e.g.: blockchains)?
- Insourcing or outsourcing?
- Explicit modeling or data-driven pragmatics?
- Service robotics for maintenance?



Much more on the upcoming book:

M. Albano, E. Jantunen, G. Papa, U. Zurutuza, "The MANTIS book: Cyber Physical System Based Proactive Maintenance", River Publishers ed., July 2018

... and it will probably be open access!!!

