

# Advanced Boiler Tube Leakage Detection using AI

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# Avoid costly repairs and unnecessary downtime

- There are between 2 and 10 tube leakages per unit and year
- Unexpected leakages drive steam generator unavailability (60% of boiler outages are due to tube leaks)
- Smaller leakages may easily cause secondary damages of costly pressure components and may lead to deformation of the entire boiler.
- A single hour of unavailability of a 500 MW unit can cause losses of \$15,000 to \$25,000.

Small leakage occurred

Underlying tube was already severely damaged

The third and fourth tube already show traces of steam leakage.

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# Acoustic leakage detection

- Equipment:
- 15 to 30 Sound Senors
- Cables
- Evaluation Computer

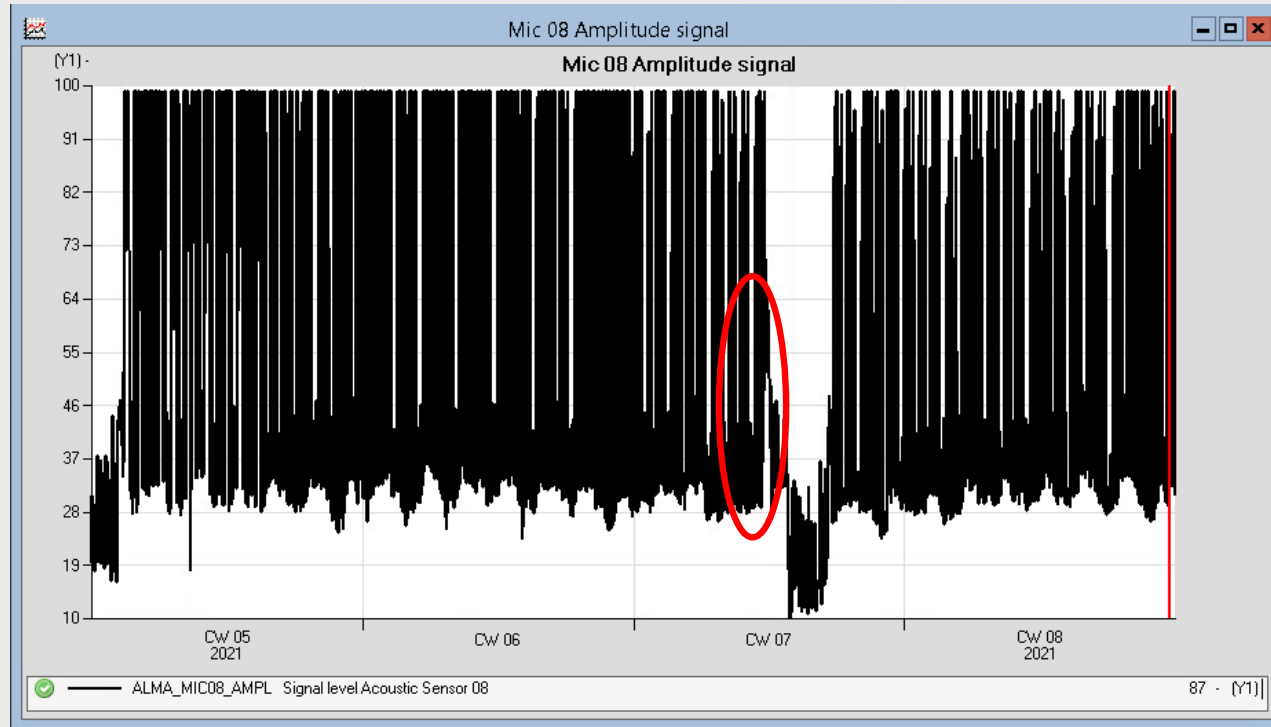


structure-borne sound sensor

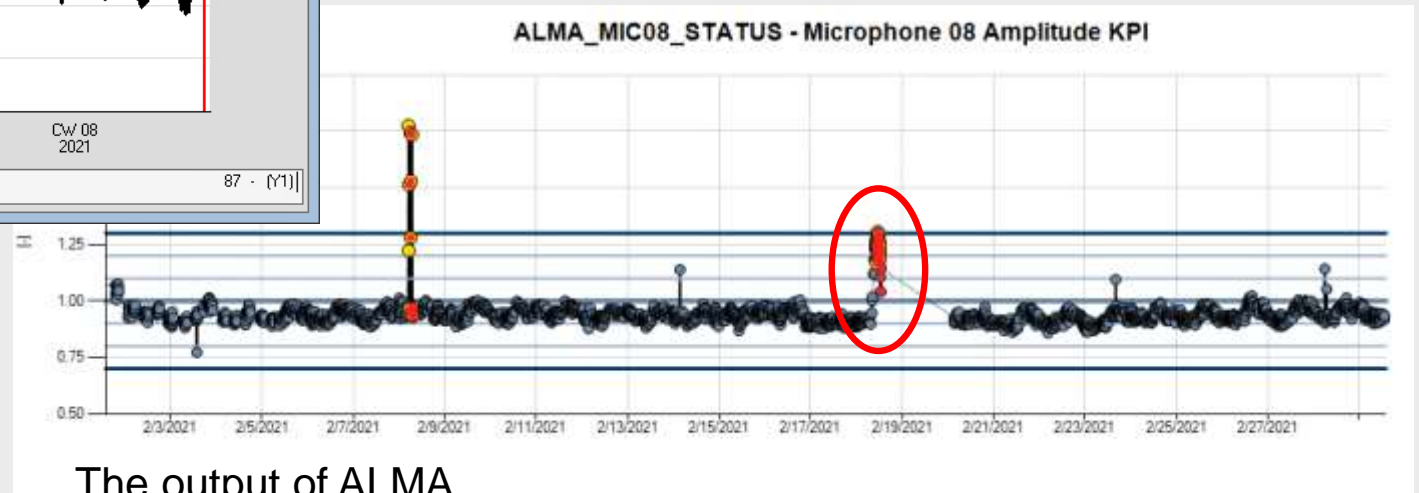
airborne sound sensor



# Do you see the leakage?



Data from the microphone

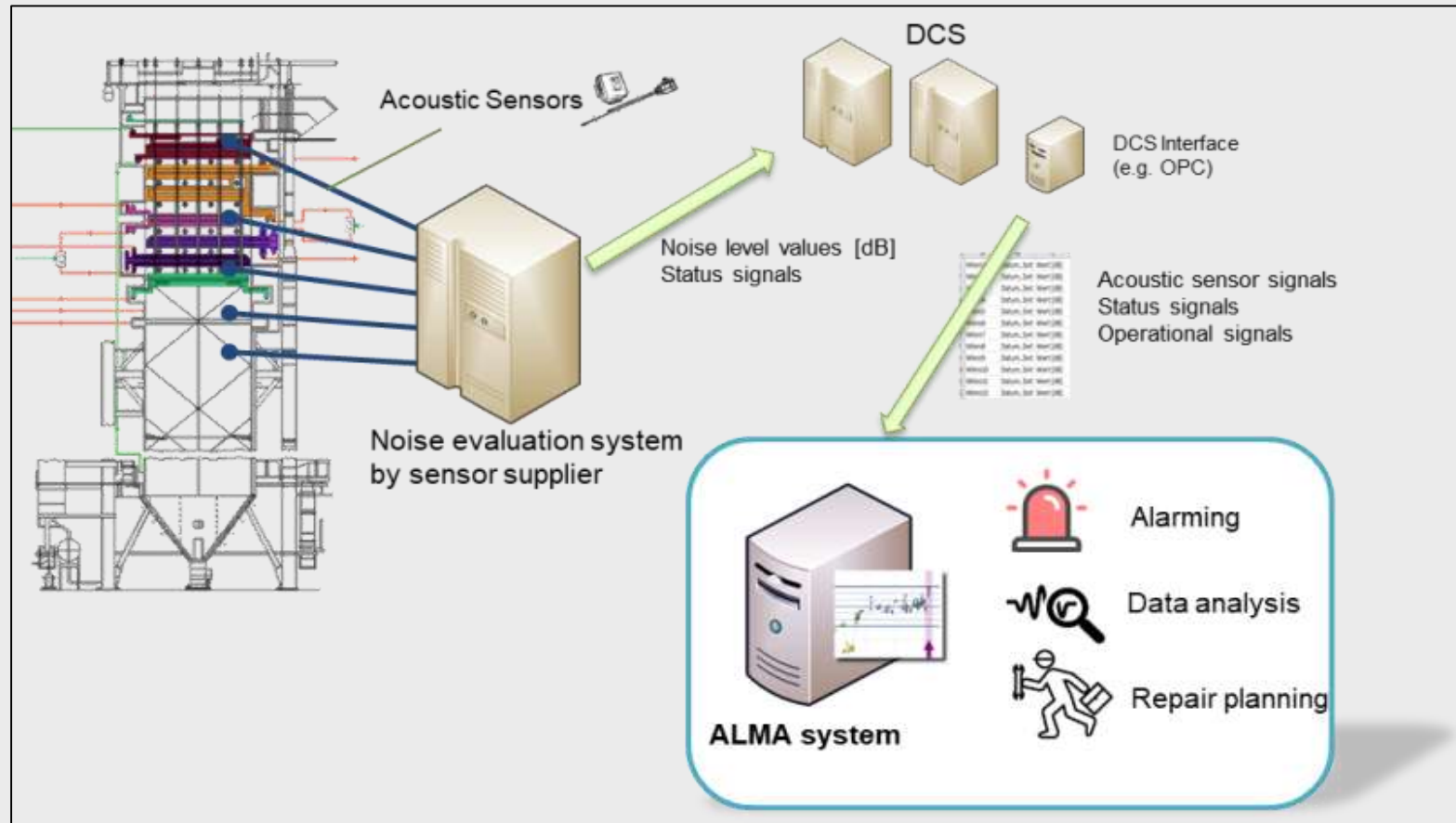


The output of ALMA



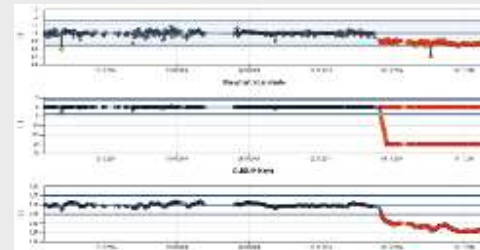
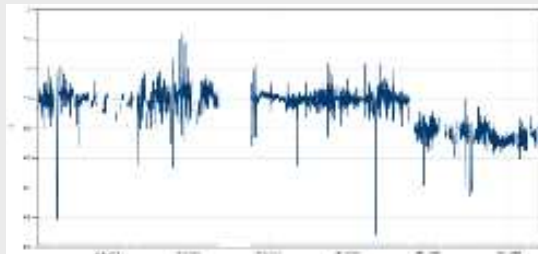
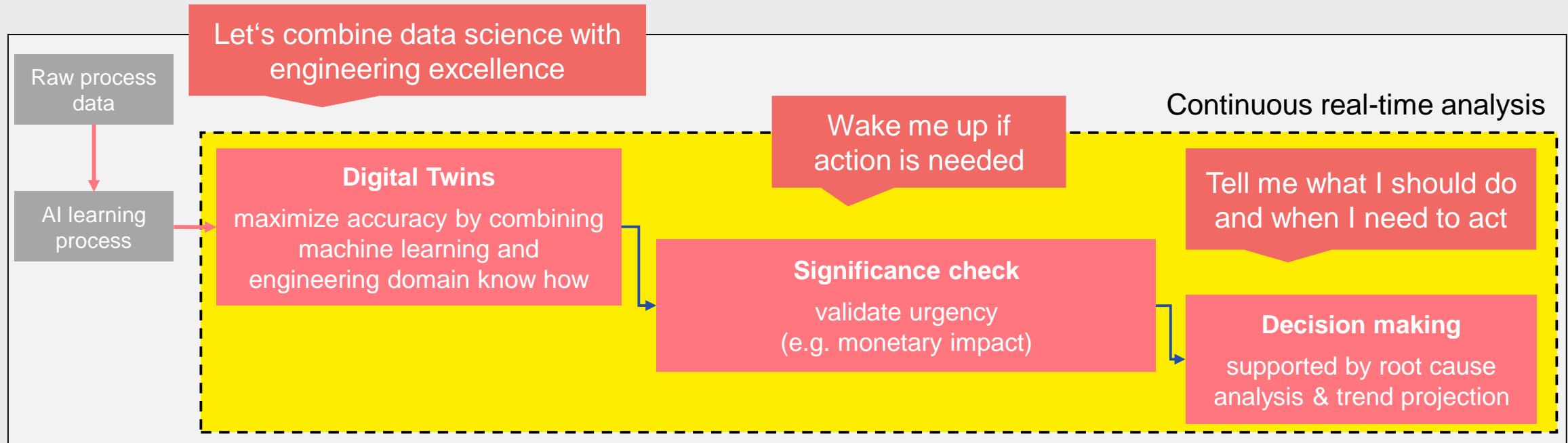
# How to make tube leakage detection fast and reliable?

- ALMA = Advanced Leakage Monitoring & Alerting



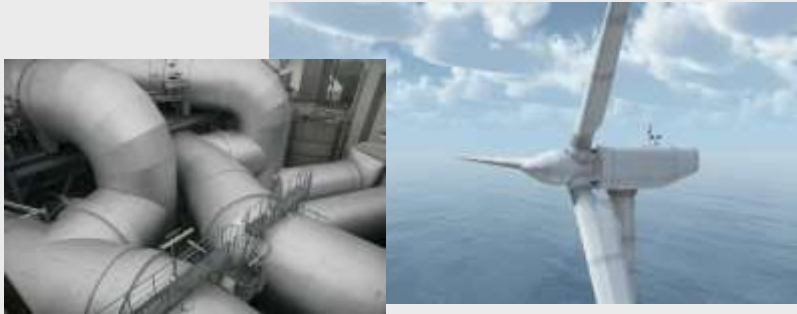
# Focus on what matters

## • The Next Level of Prediction

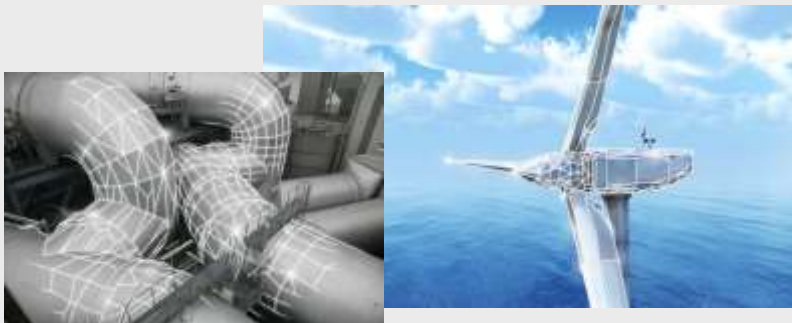
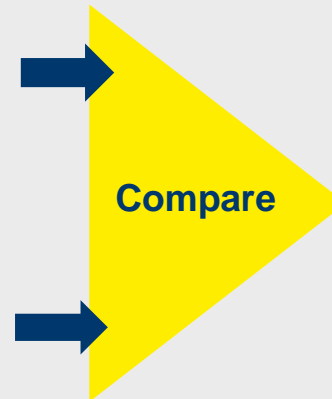


# AI to support Predictive Maintenance

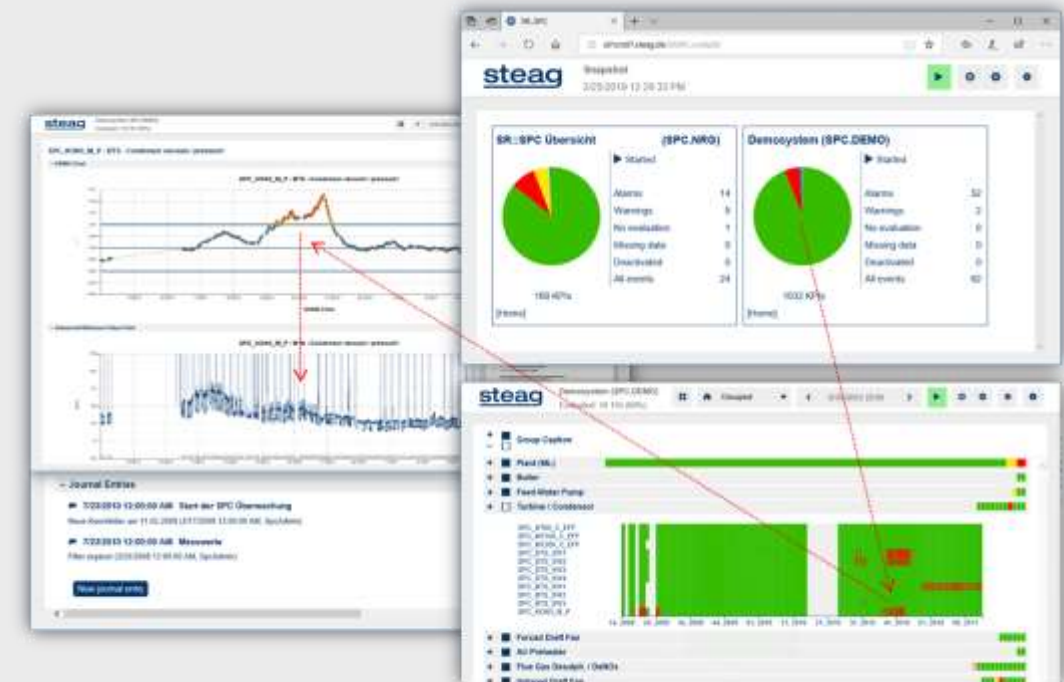
**Act on right moment, reduce additional costs of unplanned outages and increase overall planned availability through predictive maintenance**



Current sensor-based data



Digital replica / digital twin

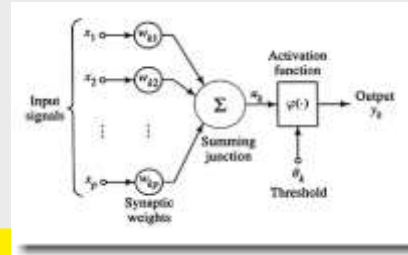


**... anomaly detection**  
**... early stage information**  
**... avoid false alarms**

# Building Digital Twins

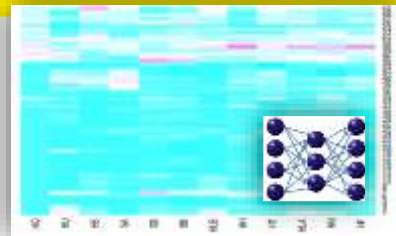


Sensor-based data

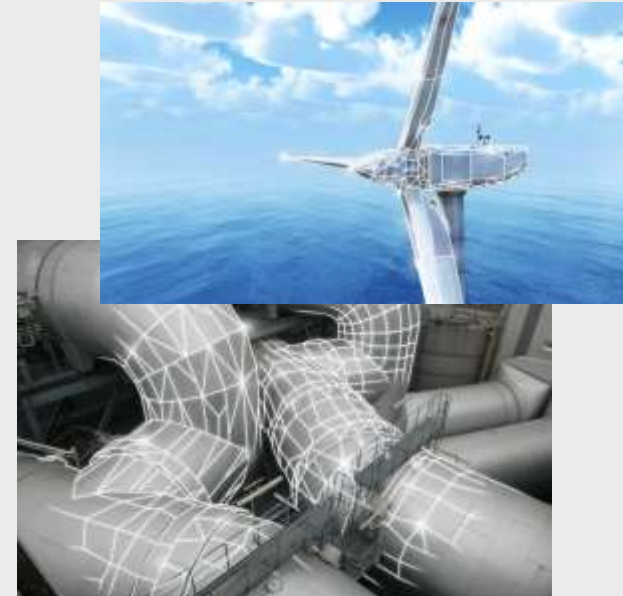


HQ KPI

Training



Big Data  
methods



Expertise on cause and effect  
is used.

Correlations between the measured values are recognized and  
independent key variables are automatically identified.

Engineering in the modeling phase enables the  
best possible performance

AI enables automatic monitoring for a large number of  
measurements

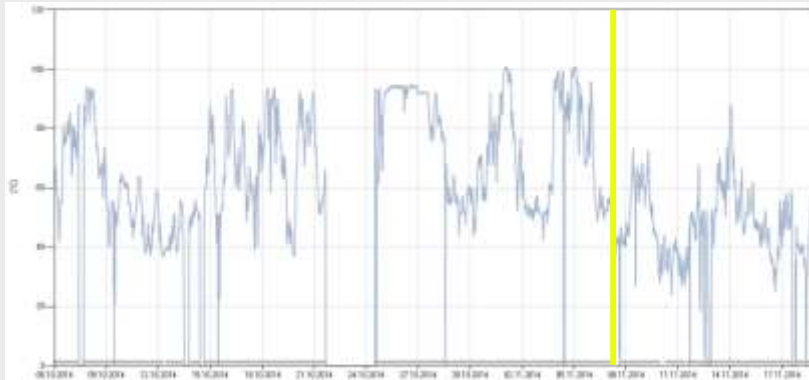
➔ “Experts HQ-KPI”

➔ “Big Data” ML-KPI

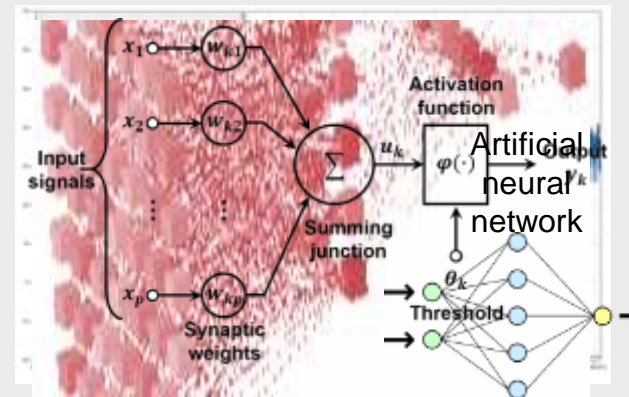


# Machine Learning and Significance Testing for Reliability

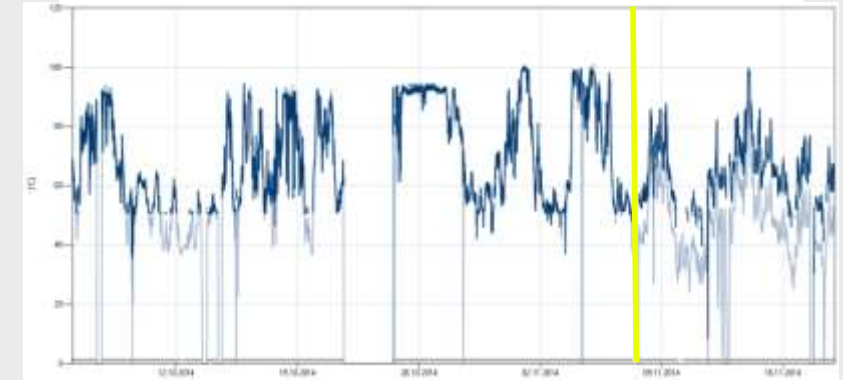
## 1. Sensor data



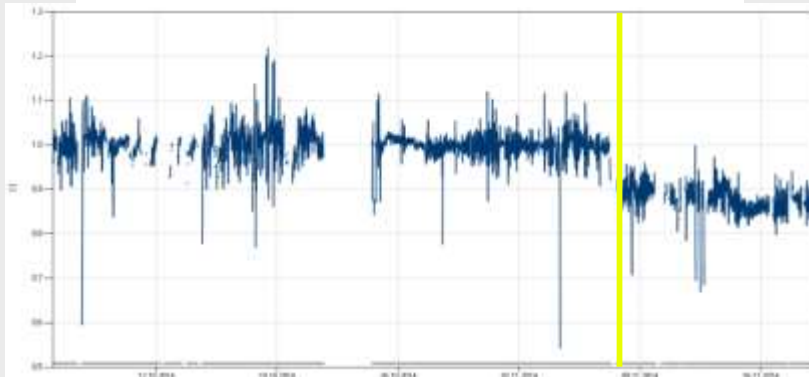
## 2. Learning process



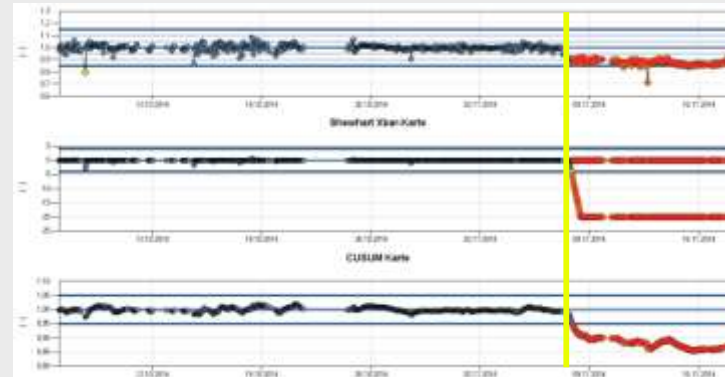
## 3. Actual and reference data



## 4. KPI = (actual/reference)



## 5. Significance test



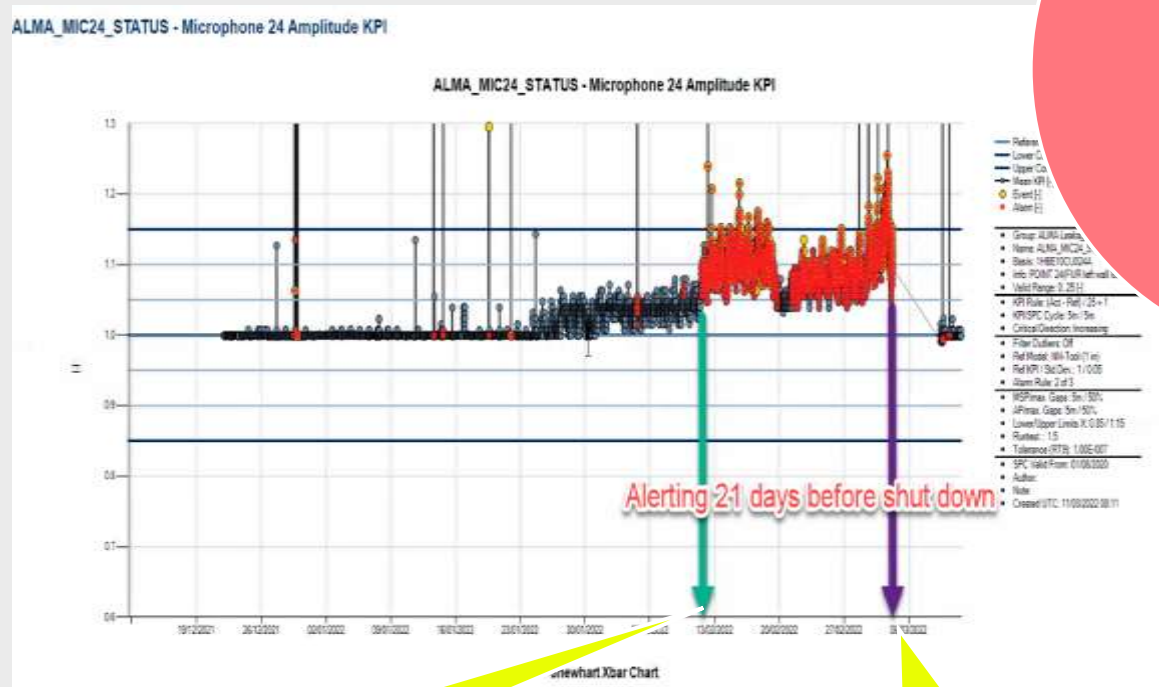
## 6. Root cause analysis



# Case study ALMA: Damages could have been avoided way before it occurred

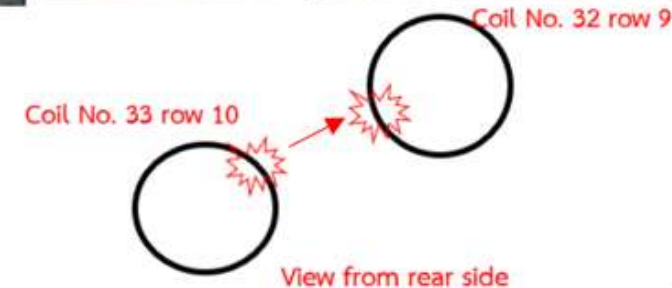
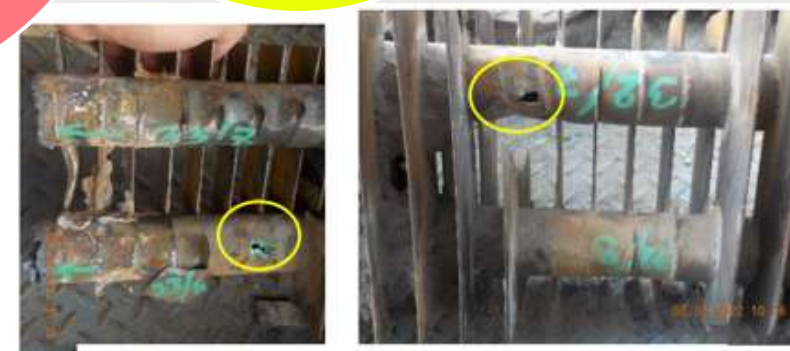
Alerted  
**21 days**  
before boiler  
shutdown

Secondary  
damages could  
have been  
avoided



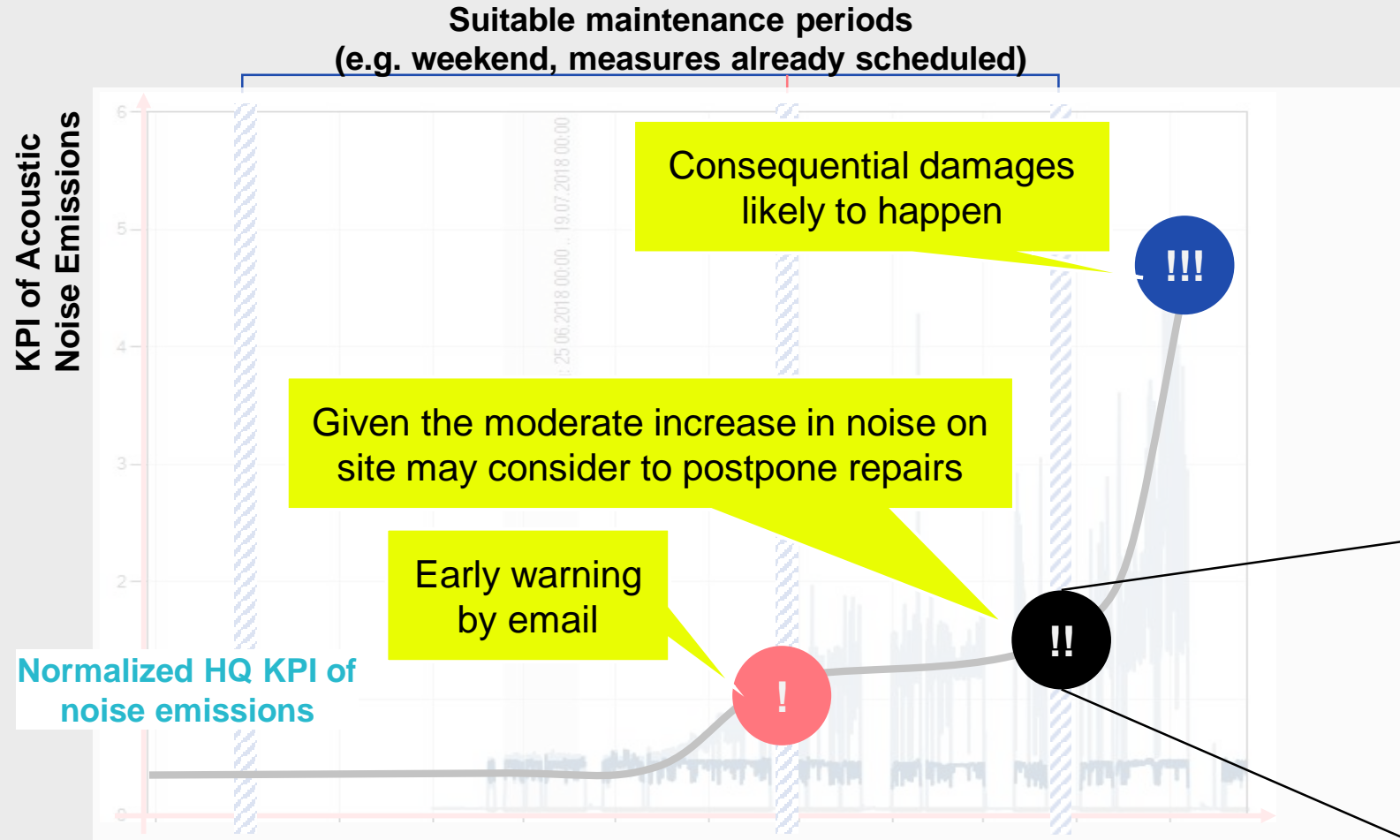
Alarm about  
leakage detection

Boiler shutdown



# Case Study: Advanced Boiler Tube Leakage Detection

- Smart planning enabled by early and accurate prediction



## Smart maintenance planning to avoid unnecessary downtime:

- Use of already scheduled maintenance events to carry out inspections and tube repairs
- Avoid consequential / secondary damages and thus shorten downtime and repair costs
- Proactive dispatch management
- Balance technical and economic impact





# Summary



- **Reliable early detection of leaking spots (no false alarms)** enabling immediate (usually less costly) measures
- **Shorter and better planned downtimes** (consequential damages can be avoided) boosting the overall viability and reliability of the asset
- **Systematic trend analysis** telling the site team when repairs are to be scheduled (excellent decision making support)
- **Scalable platform solution** that grows flexibly with the individual requirements the client may have
- Long-term IT solution partner with a proven track record ensuring the **sustainability of the investment**



Thank you for your  
attention.

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