

# Successful implementation of a customized predictive maintenance solution in industrial manufacturing

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**anacision**  
an **EXXETA** company

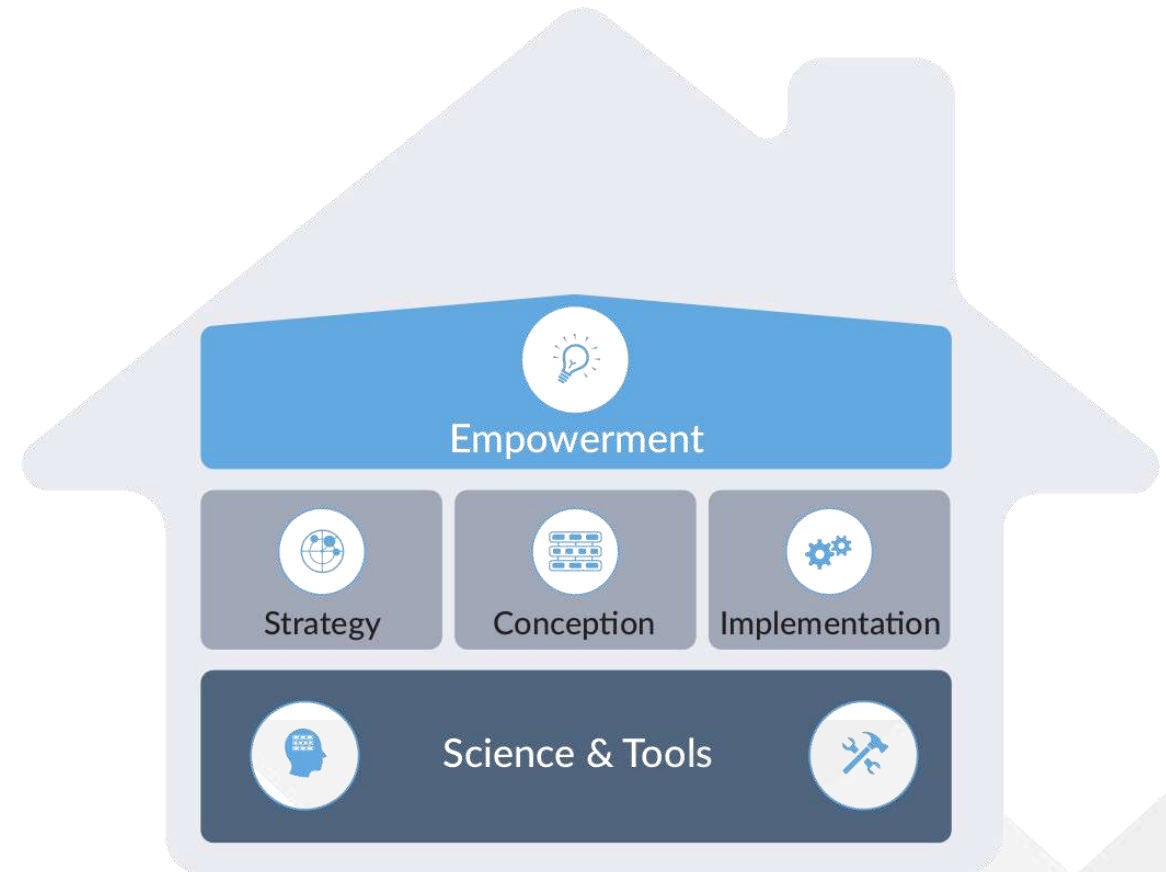


# The mission of anacision

[,æneə' sɪʒn] analytics + decision

## Our mission

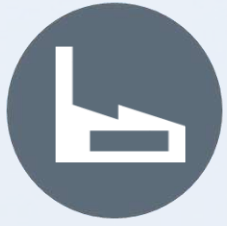
We help to make better decisions  
with intelligent data analysis.











**Hundreds of plants**



**More than 500.000 Sensors**



**Reactive, time based and  
condition based maintenance**



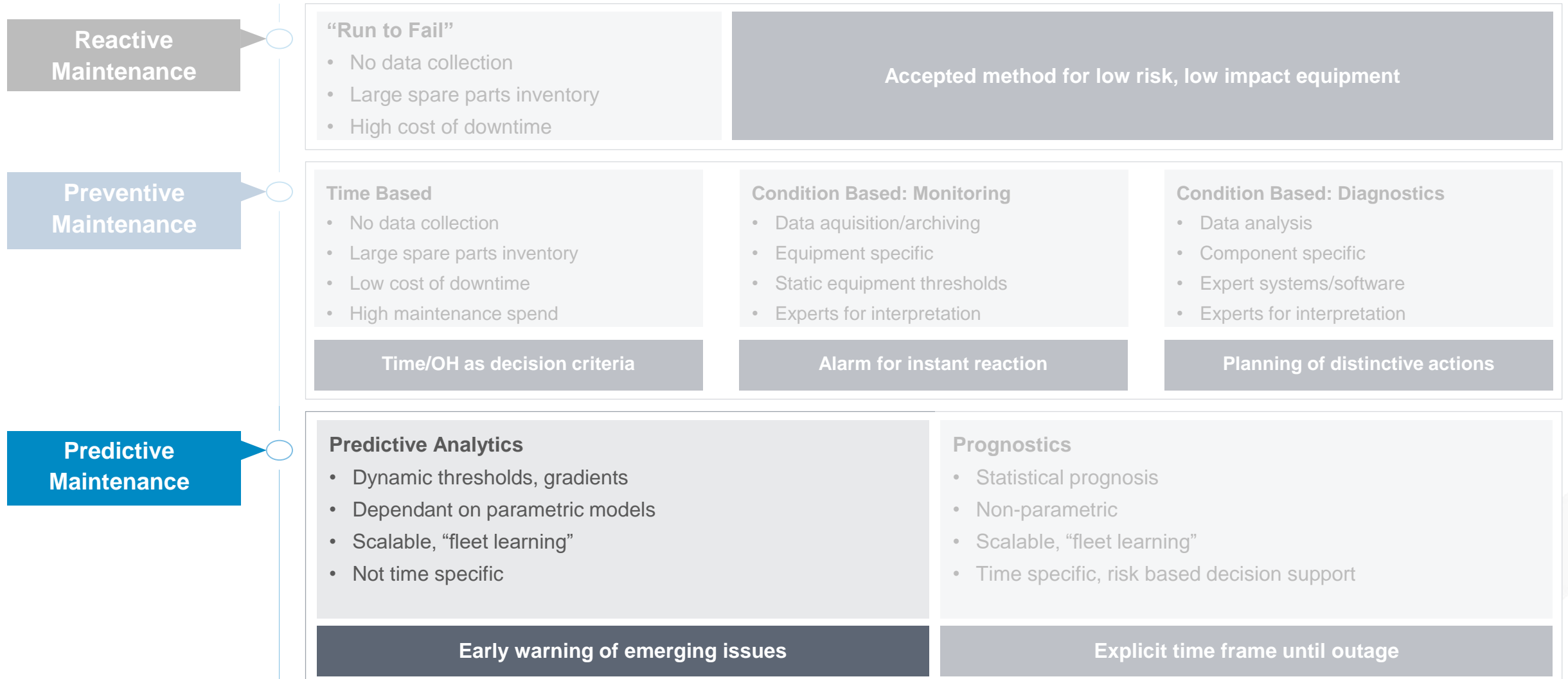
**Failures rare but expensive**



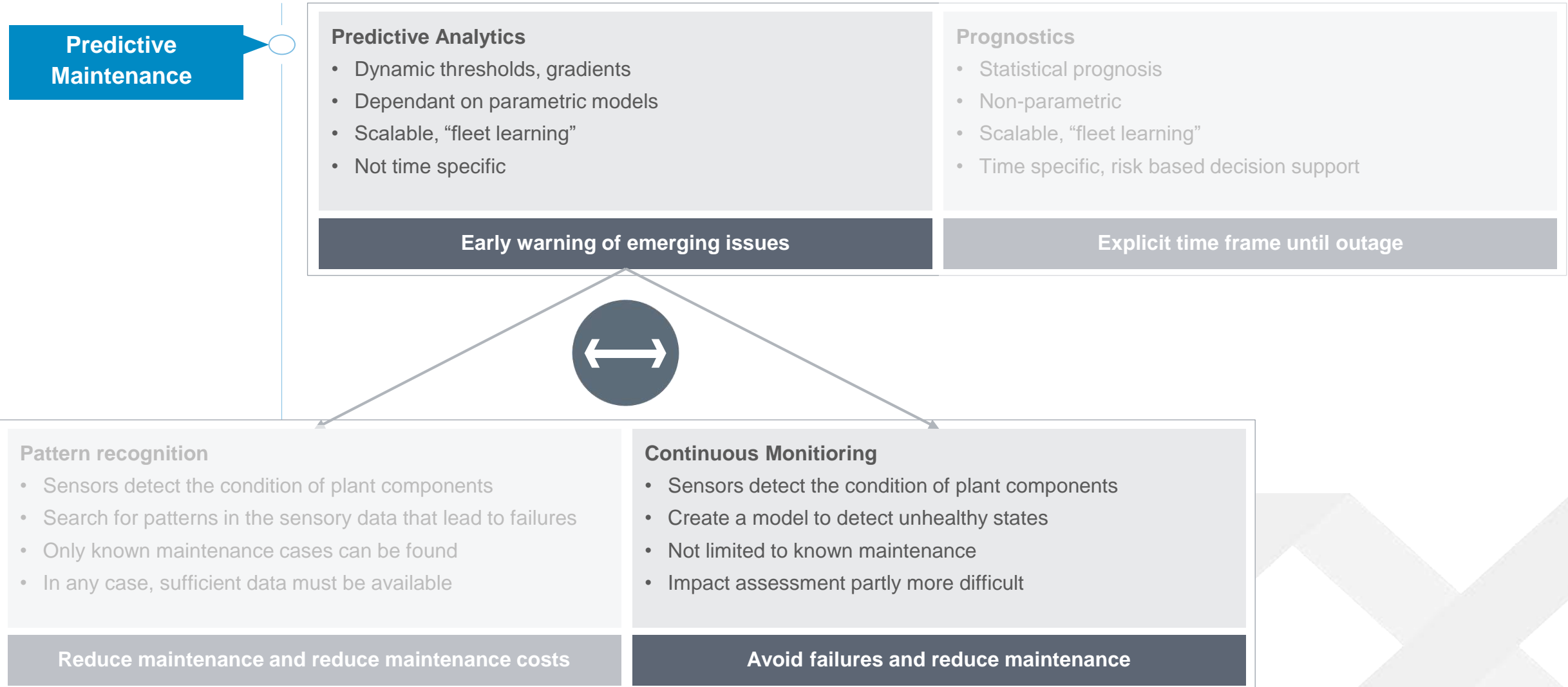
**Every day GBs of sensor data  
are collected centrally**

**Predictive Maintenance: Predict necessary maintenance activities  
and avoid unplanned shutdowns**

# Predictive Maintenance Strategic Approach



# Predictive Maintenance Strategic Approach



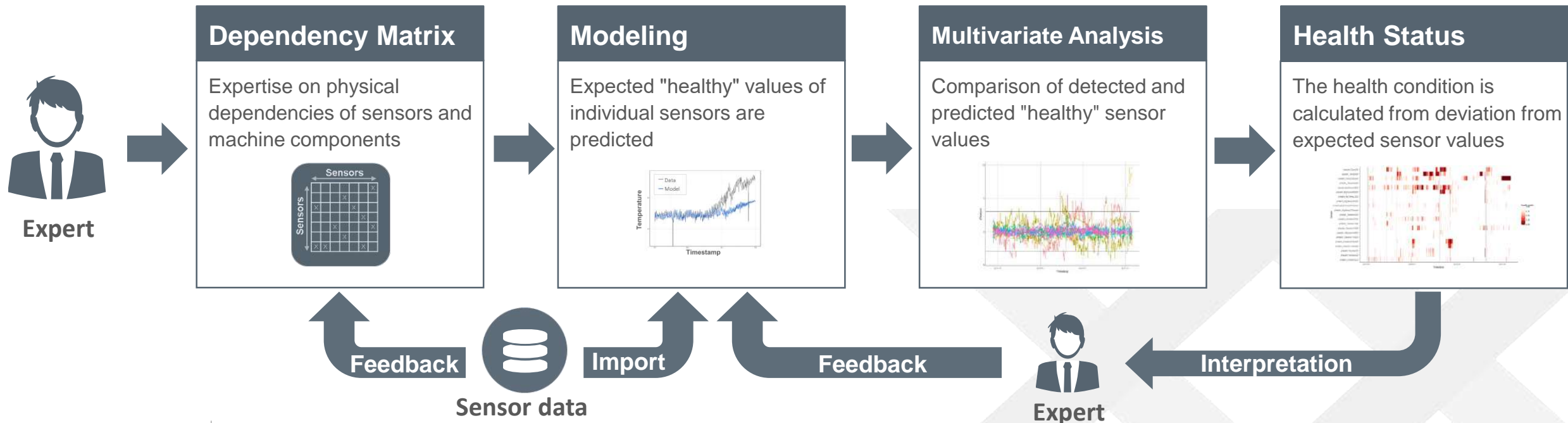
# Our approach to continuous monitoring of the health of industrial plants

## Goals

- Continuous measurement and interpretation of the health of industrial plants
- Automated early indicators for unhealthy developments
- Reduced maintenance costs and reduced plant downtimes

## Approach

- Combination of expert knowledge and sensor data to predict unhealthy states



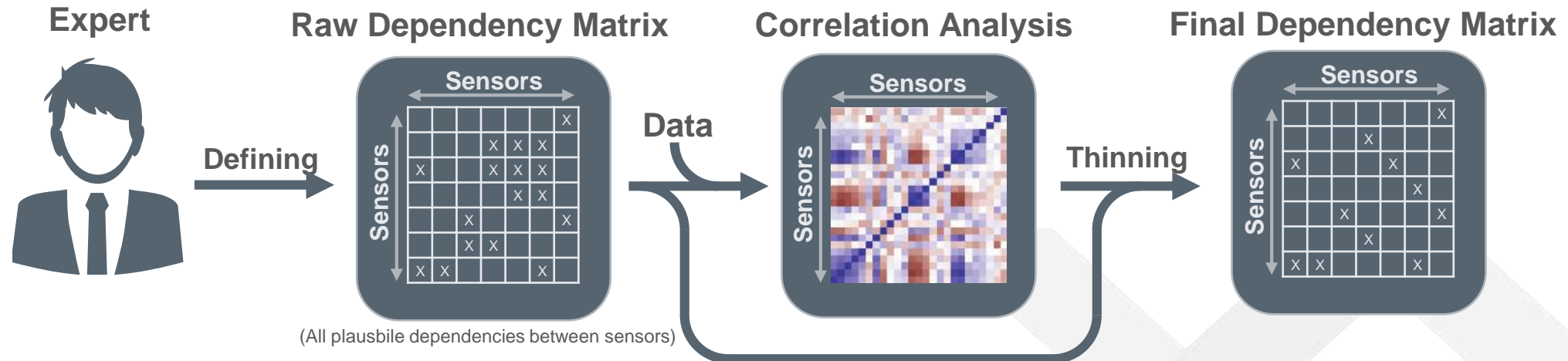
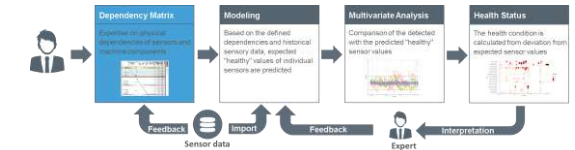


# From expert knowledge to dependency models

// A high number of potential correlations were identified, but

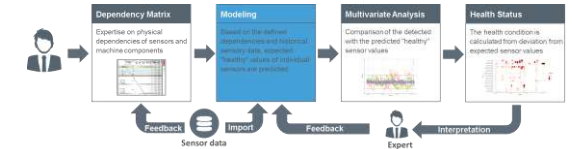
- some correlations can be spurious correlations resulting from a shared influence
- statistical models can be unreliable in case of many highly correlated predictors (multicollinearity)

→ The dependency matrix has to be thinned out in order to generate reliable and expressive models

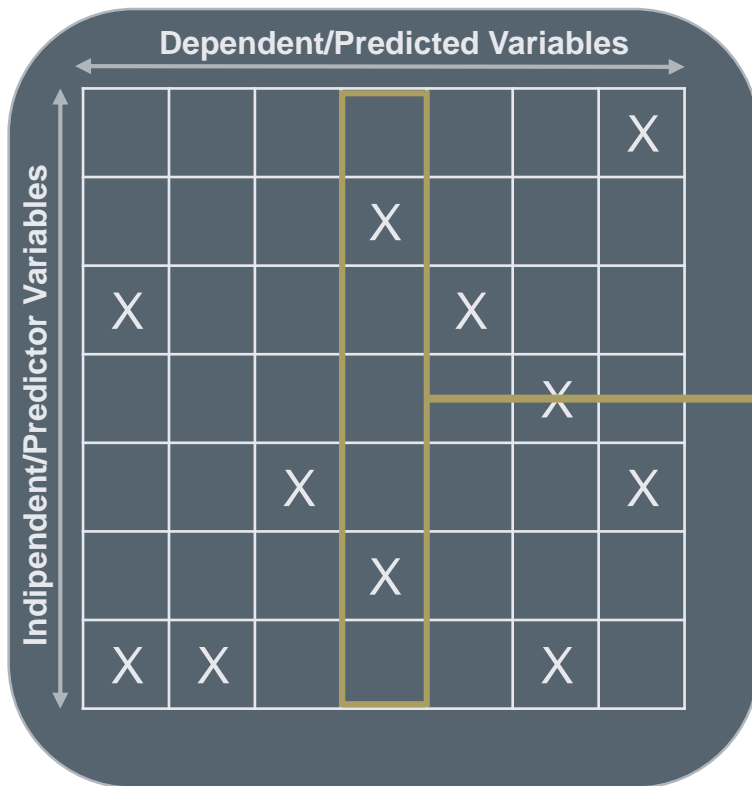




# From influence matrix to linear regression models



// Models are generated dynamically based on defined dependencies in the influence matrices



// A linear regression model for the variable Y is defined as

$$Y = \beta_0 + \sum_{i=1..p} \beta_i X_i$$

with predictor variables  $X_i$ , constant  $\beta_0$  and coefficients  $\beta_i$  for all  $p$  predictor variables ("X" in dependency matrix)

// Example: Temperature

$$Temp = \beta_0$$

$$+ \beta_{pressure} \cdot Pressure$$

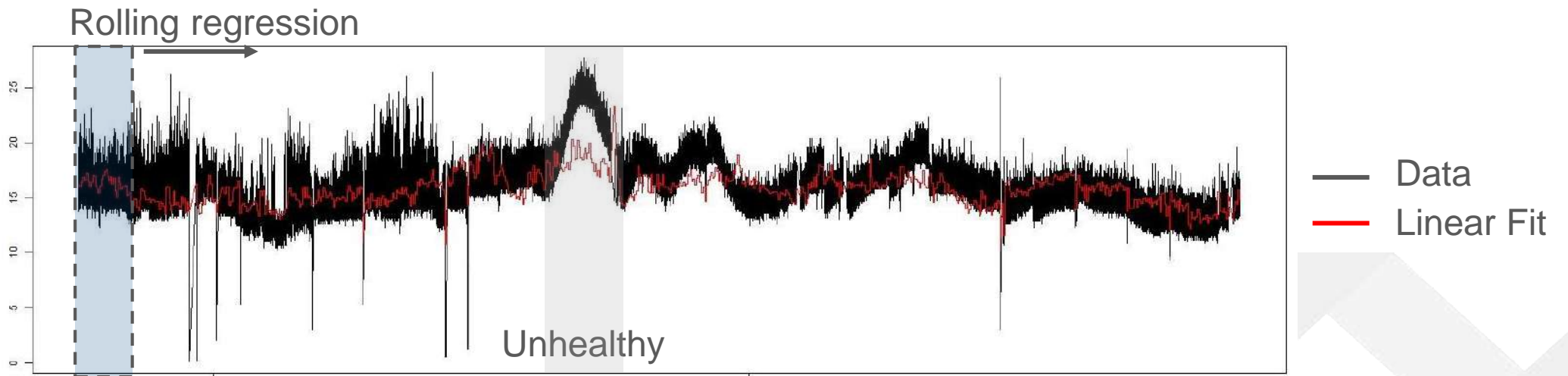
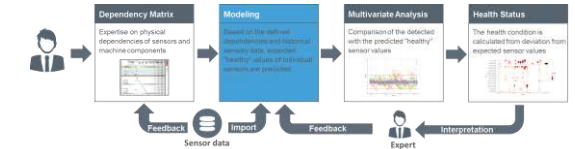
$$+ \beta_{Ambient\ Temperature} \cdot Ambient\ Temperature$$

# Linear regression model representing „healthy“

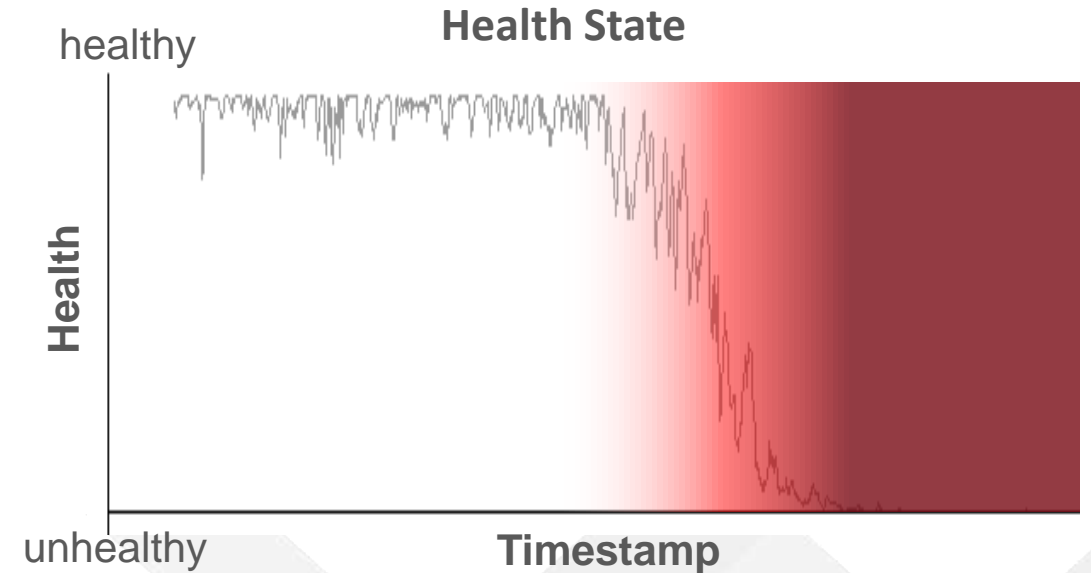
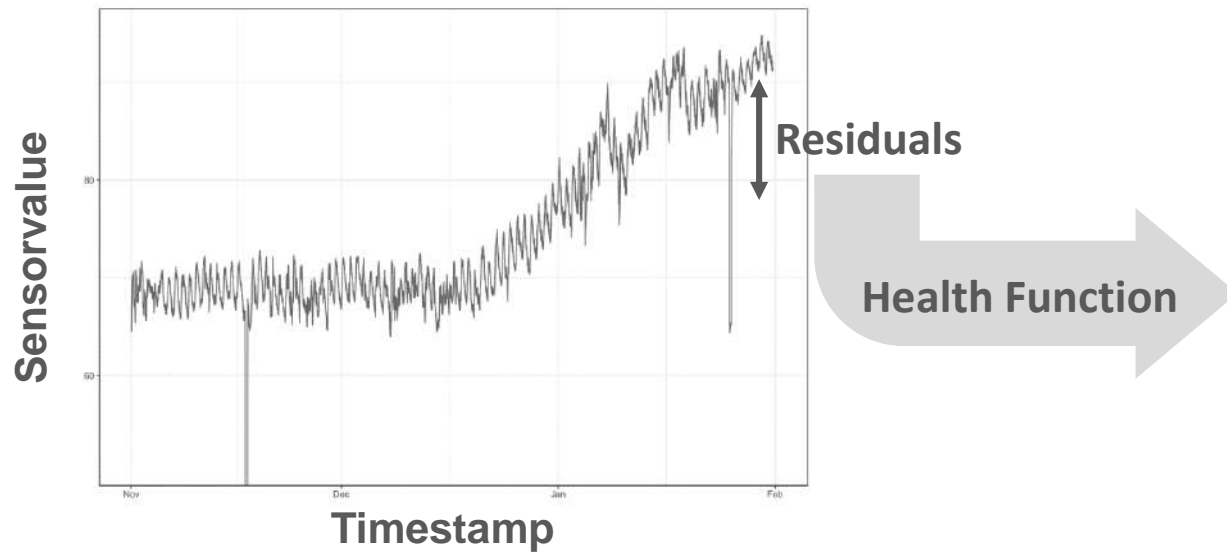
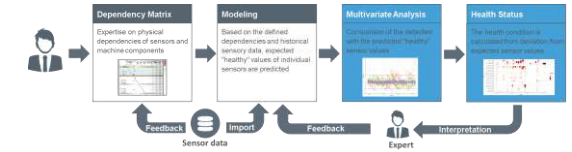
Linear Regression Model

Anomaly Detection

Rolling Regression

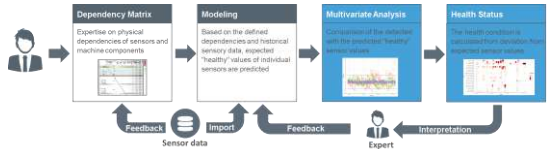
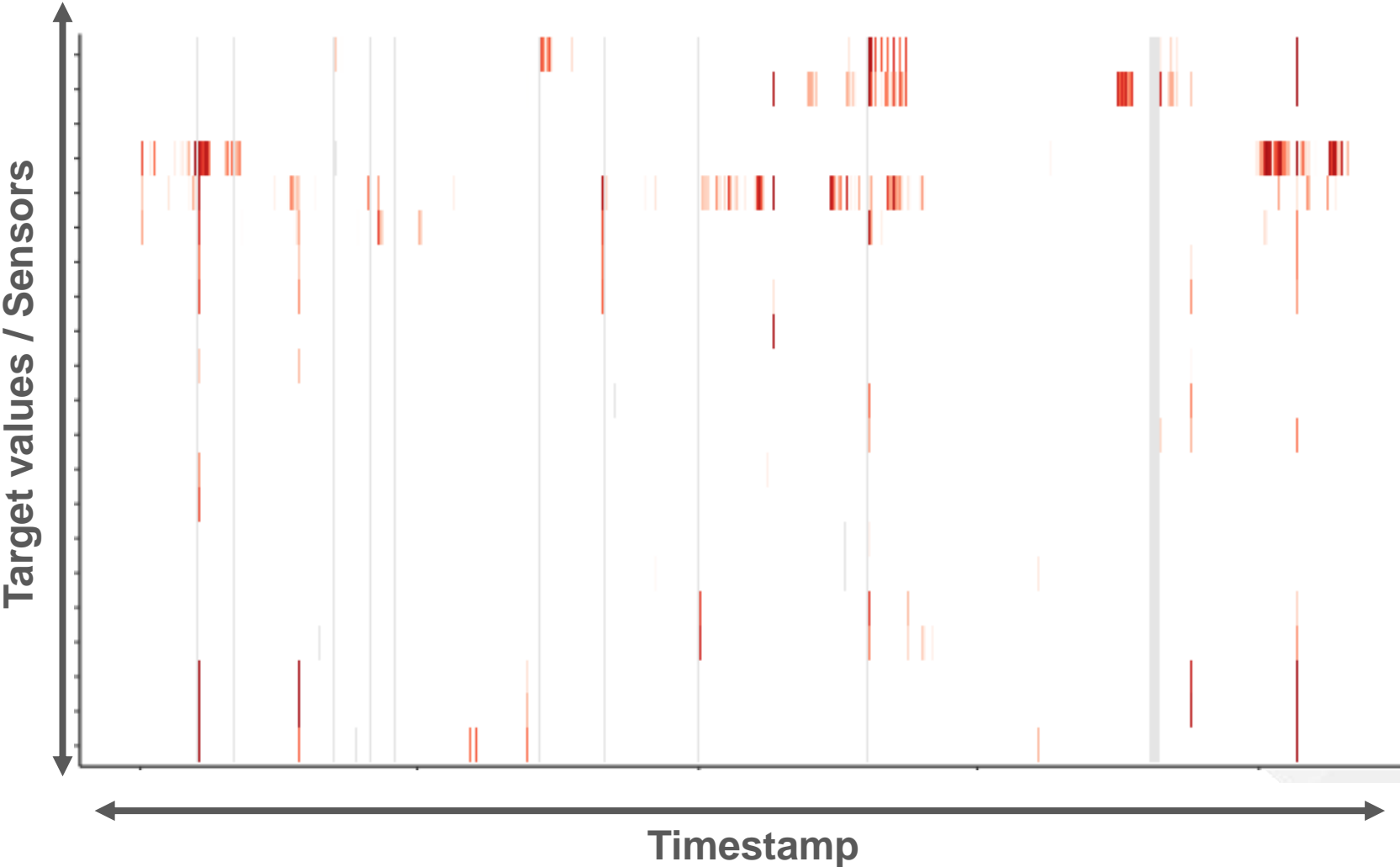


# From real-time sensory data to continuous evaluation of health state





# Overview of the health status of one plant

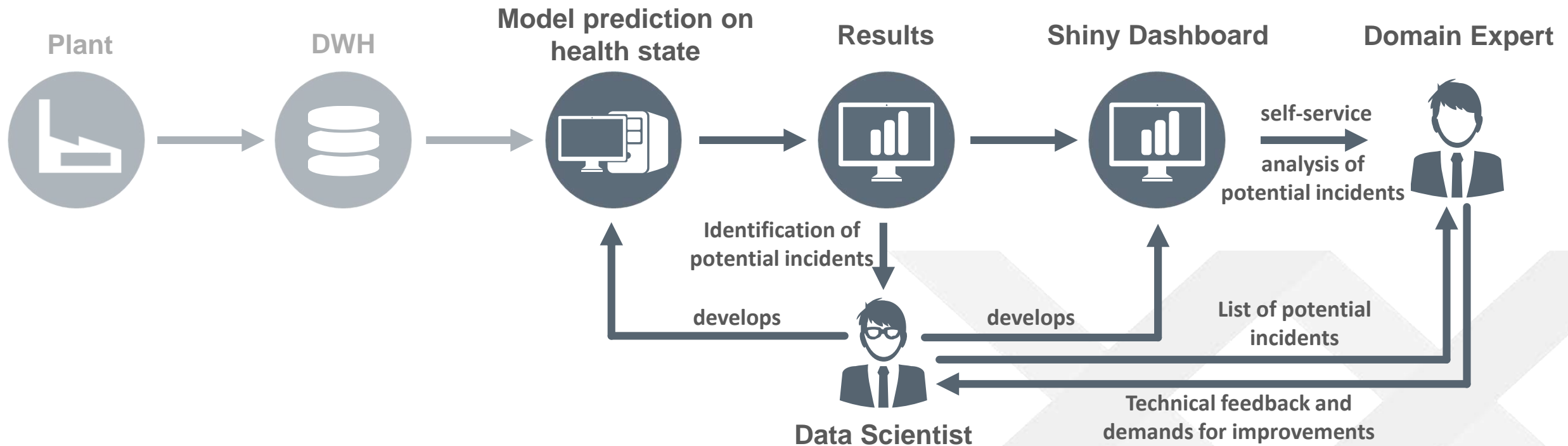


# Project setup and lessons learned

Customer has analyzed a large number of standard solutions available on the market without success.

## Key success factor:

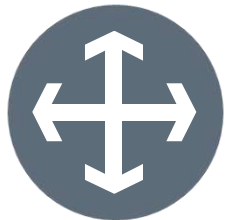
Customized Solution with continuous exchange between data scientists (anacision GmbH) and domain experts (customer).



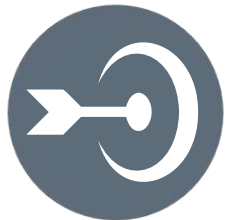
# Summary



Continuous monitoring can provide elementary advantages over pure pattern-based procedures



Easy to transfer to other machines and industries



Proofed in various projects, among others in a German DAX Company



# Contact person



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