



# **Combined physical and AI-based predictive maintenance for components**

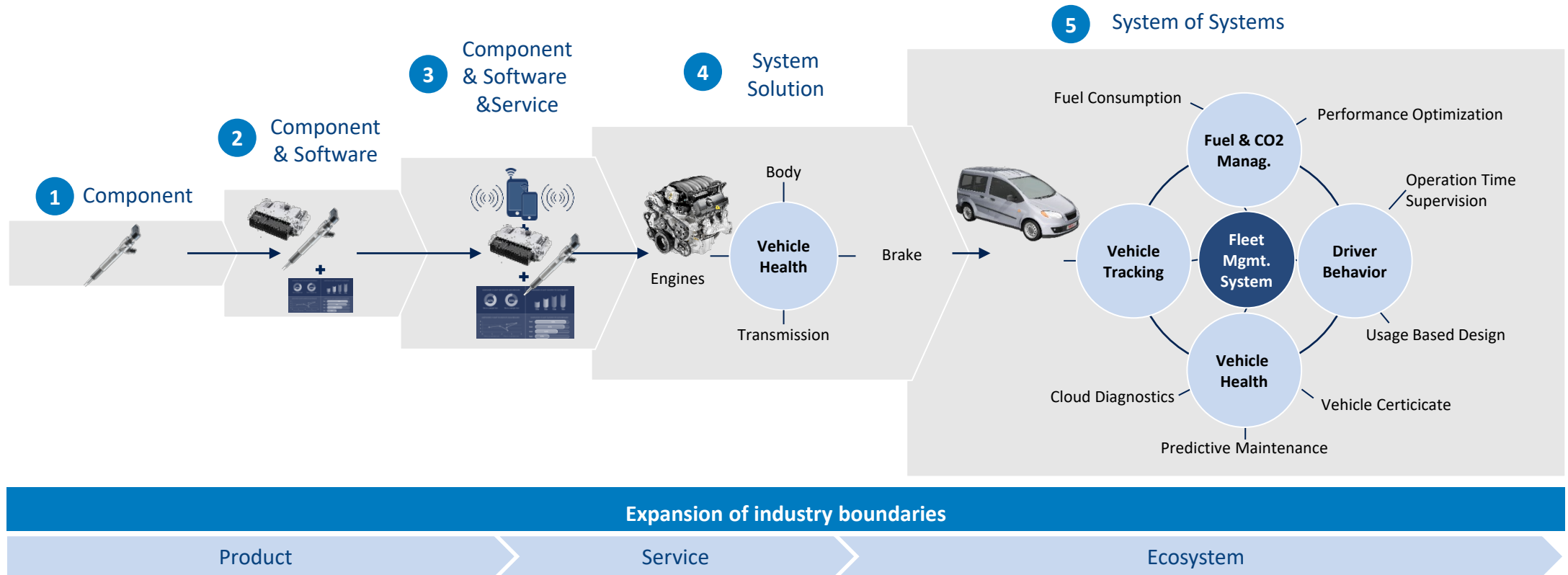
Dr. Walter Lehle



# Predictive Maintenance

# Predictive Maintenance for Components

## Services - intelligent components



# Predictive Maintenance for Components

## Predictive Diagnosis - avoid unplanned repair

Upcoming failure during operation

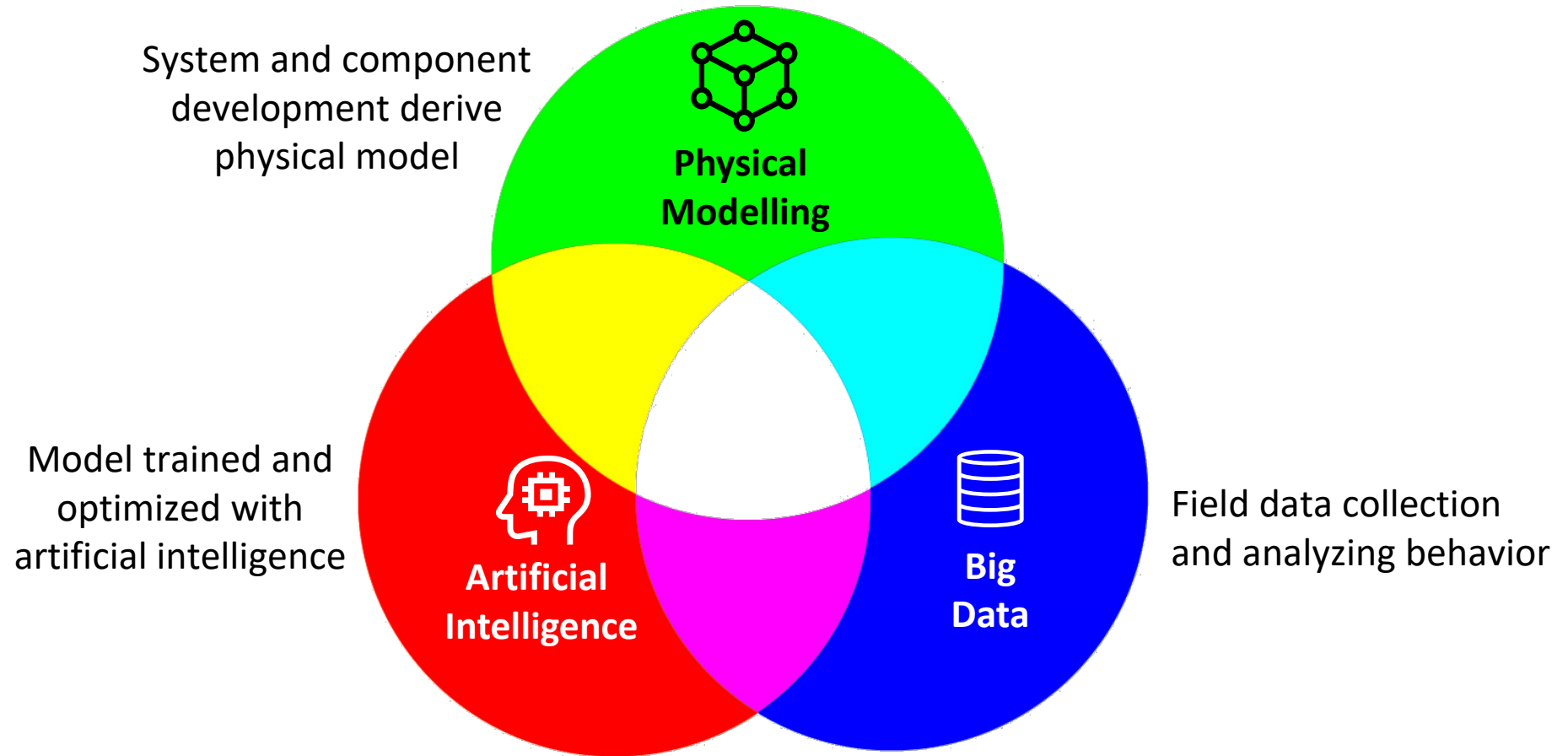


**Avoid breakdown - increase uptime**  
**Identify critical components**

**Optimize maintenance - reduce repair visit**  
**Determine components individual RuL**

# Predictive Maintenance for Components

## **Solution** - options versus combination



# Predictive Maintenance for Components

## AI outlook - classification example



### How to pick up only the sweet oranges ?

→ Differentiate between two classes: “sweet” & “not sweet”

1. Learn from experience: design diagnostic features

- Color**
- Size**
- Softness**
- Land of origin**
- Season**

2. Develop an algorithm:

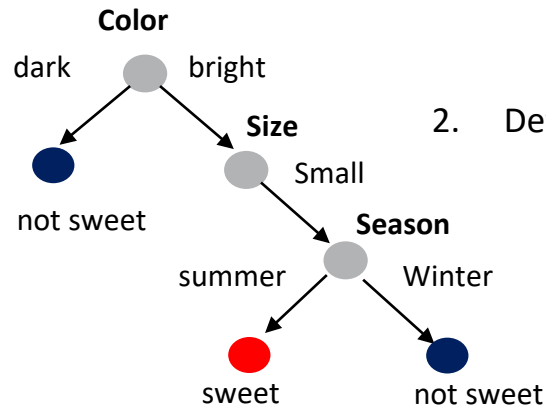
*If* **Color** == Dark:

**Size** == Small:

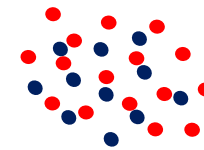
*if* **Season** == Winter

.....

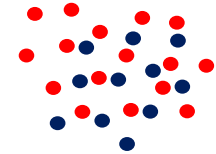
*then*: Orange is not sweet



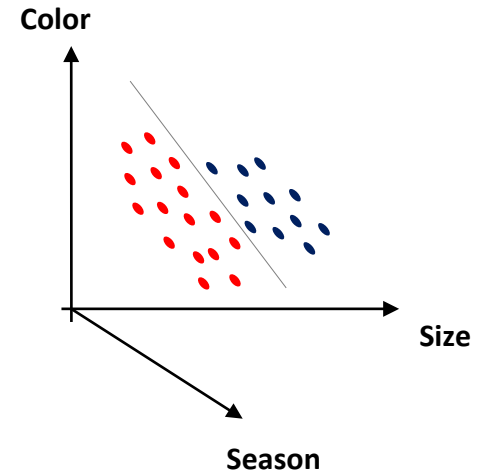
Color/ Season



Season/ Size



**BUT! Separable in 3D:**



### Main task of AI:

Automatically determine correlations between diagnostic features (color, size, season etc.) and class (sweet or not sweet) in a multidimensional space.

If a system is complex, it is intractable to analyse the complete search space manually

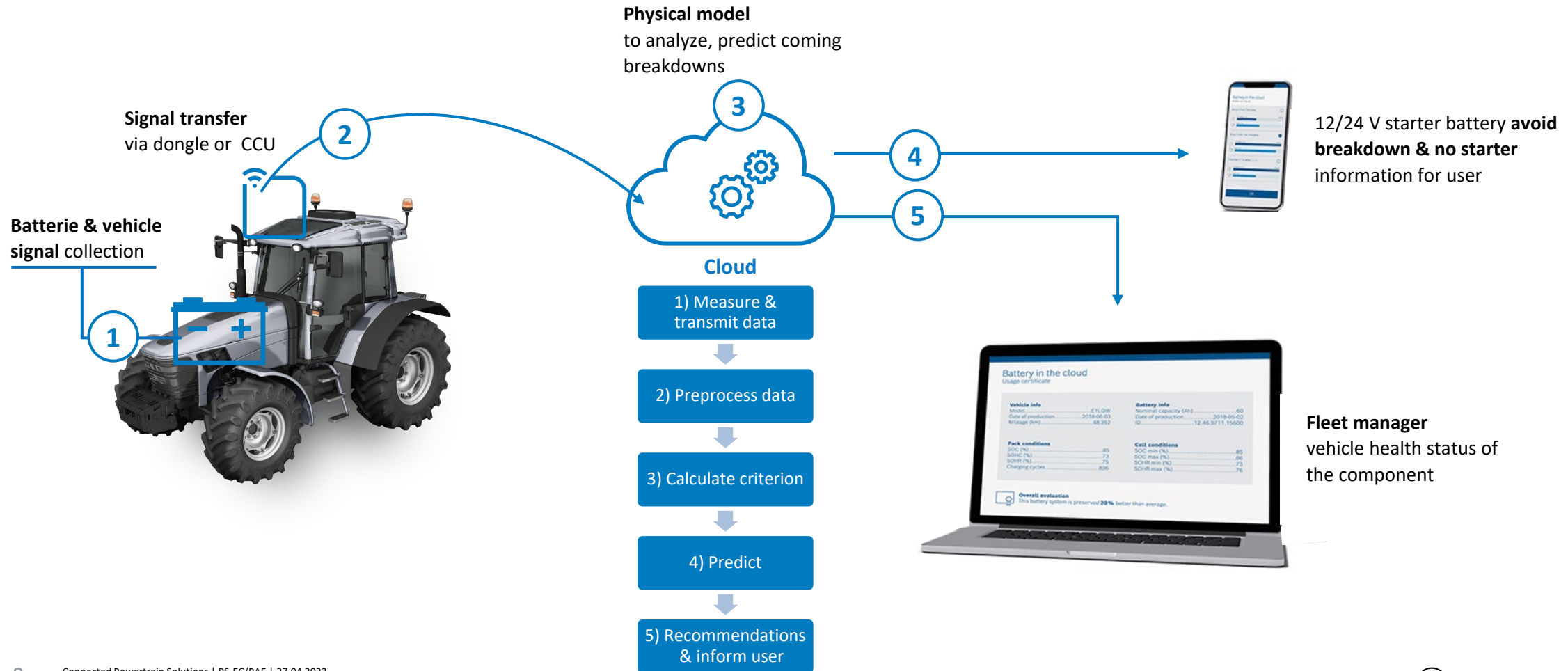


# Example Batteries

# Predictive Maintenance for Components

## 12V/24V Batteries - How it works

Avoid breakdown - increase uptime  
Identify critical components





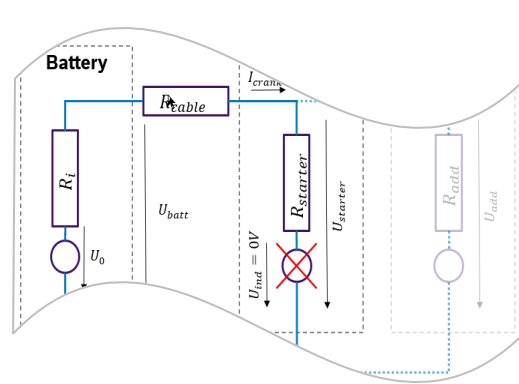
# Predictive Maintenance for Components

## 12V/24V Batteries – Physical Based Approaches add AI

Avoid breakdown - increase uptime  
Identify critical components

### 1 Physical electric model

- Proprietary algorithms to describe system behavior
- Balanced assumptions and approximations



### 2 Key parameter

Signal Name	Values
Voltage	In „ms“
Vehicle Status	On / Off
Distance	Odometer

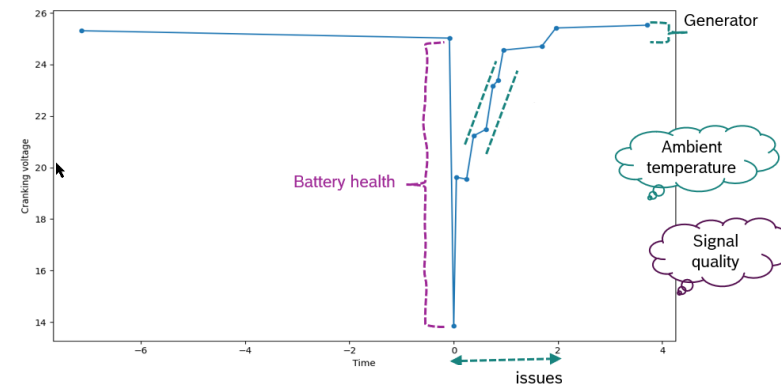
- General measured variables
- Focusing on the variables associated phase during start

### 4 Deploying Concept

- Learning about Vehicle Battery Health over time
- Categorizing the Battery Health, thus enabling the Customer to take necessary steps



### 3 Concept: Focus on Cranking Voltage

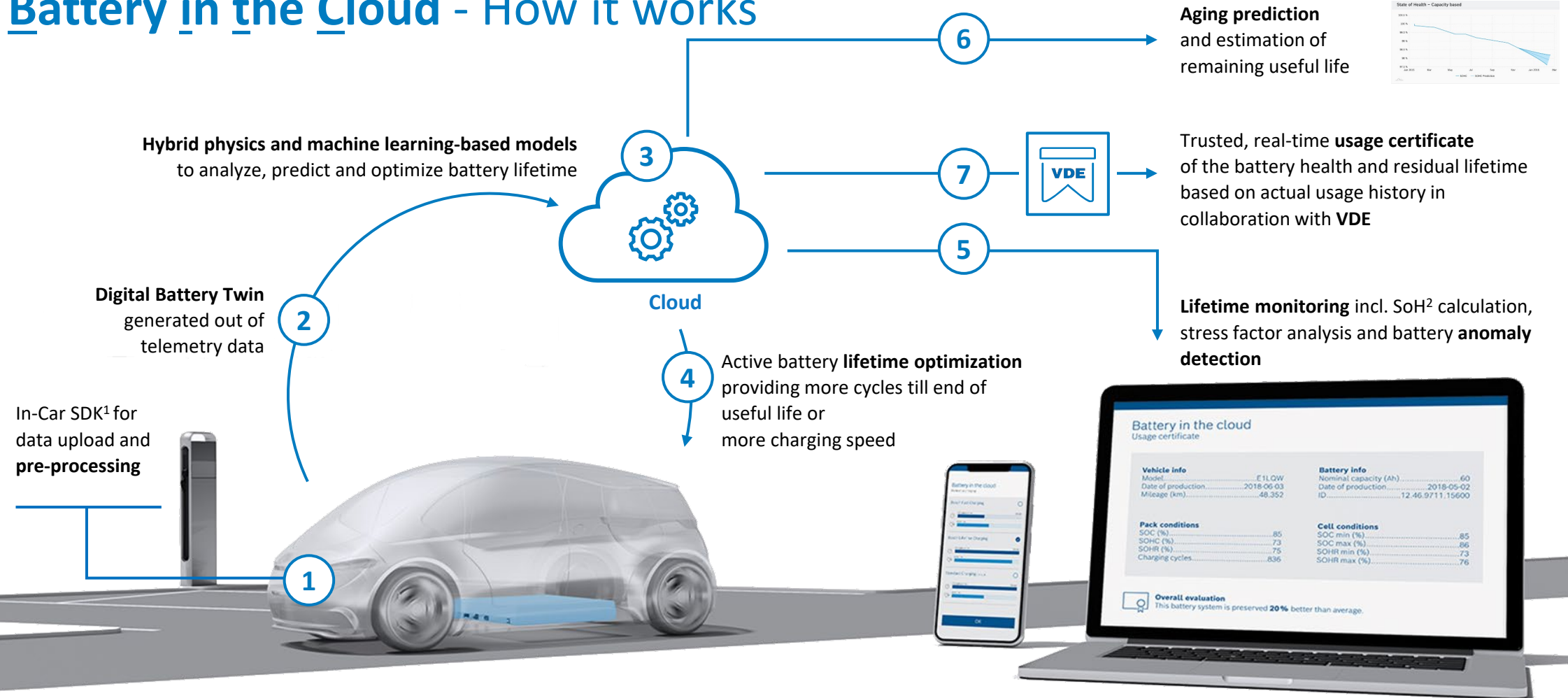


- Generate customer value by better understanding
- Improve performance or knowledge about SOH

# Predictive Maintenance for Components

## Battery in the Cloud - How it works

Optimize maintenance - reduce repair visit  
Determine components individual RuLc



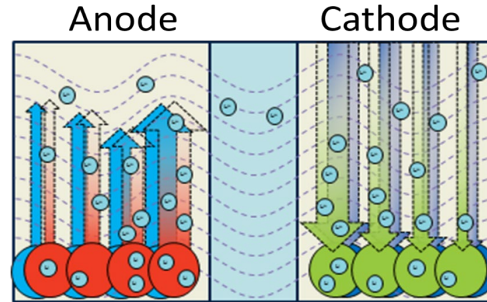
# Predictive Maintenance for Components

## BitC - Life Cycle of Model Based Approaches

Optimize maintenance - reduce repair visit  
Determine components individual RuLc

### 1 Basis: Battery electrochemical Model

- ▶ Algorithms for describe system
- ▶ Balanced assumptions and approximations
- ▶ Fusion of approaches (e.g., hybrid model)



### 2 Model Training

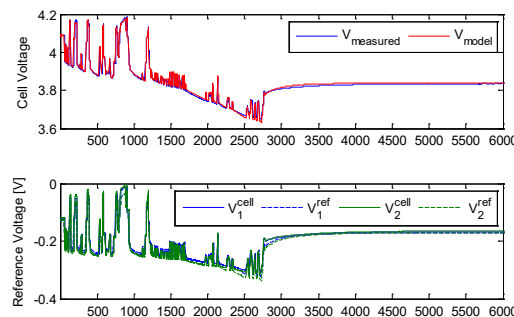
Parameter	Value	Unit	Source
Electrolyte conductivity	1.0	S/cm	Manufacturer data
Electrode thickness	0.001	m	Manufacturer data
Electrode porosity	0.4	-	Manufacturer data
Electrode active area	1.0	m²	Manufacturer data
Electrode initial state of charge	0.5	-	Manufacturer data
Electrode initial temperature	298.15	K	Manufacturer data
Electrode initial pressure	1.0	bar	Manufacturer data
Electrode initial humidity	1.0	-	Manufacturer data
Electrode initial oxygen concentration	0.21	-	Manufacturer data
Electrode initial nitrogen concentration	0.79	-	Manufacturer data
Electrode initial carbon concentration	0.0	-	Manufacturer data
Electrode initial sulfur concentration	0.0	-	Manufacturer data
Electrode initial phosphorus concentration	0.0	-	Manufacturer data
Electrode initial potassium concentration	0.0	-	Manufacturer data
Electrode initial calcium concentration	0.0	-	Manufacturer data
Electrode initial magnesium concentration	0.0	-	Manufacturer data
Electrode initial sodium concentration	0.0	-	Manufacturer data
Electrode initial bromine concentration	0.0	-	Manufacturer data
Electrode initial iodine concentration	0.0	-	Manufacturer data
Electrode initial chlorine concentration	0.0	-	Manufacturer data
Electrode initial fluorine concentration	0.0	-	Manufacturer data
Electrode initial hydrogen concentration	0.0	-	Manufacturer data
Electrode initial helium concentration	0.0	-	Manufacturer data
Electrode initial neon concentration	0.0	-	Manufacturer data
Electrode initial argon concentration	0.0	-	Manufacturer data
Electrode initial krypton concentration	0.0	-	Manufacturer data
Electrode initial xenon concentration	0.0	-	Manufacturer data
Electrode initial radon concentration	0.0	-	Manufacturer data

- ▶ Material properties
- ▶ Size and form
- ▶ Usage of the battery

### 4 Validate and optimize model

Example:

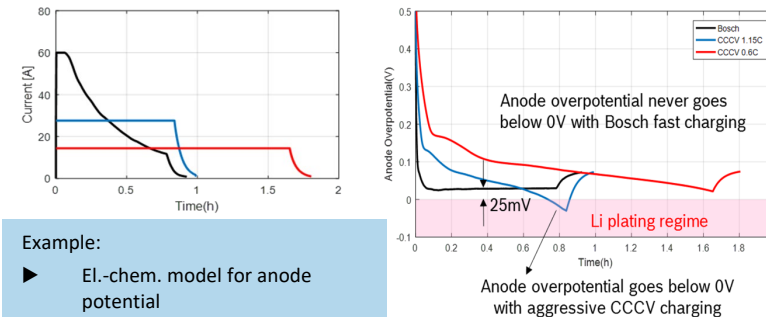
- ▶ 3-electrode measurements for model validation (lab)
- ▶ Demonstration of charging speed and SOH effects in lab and pilot fleet



### 3 Deploy model in battery services

Example:

- ▶ El.-chem. model for anode potential
- ▶ Enables shorter charging time or reduced degradation

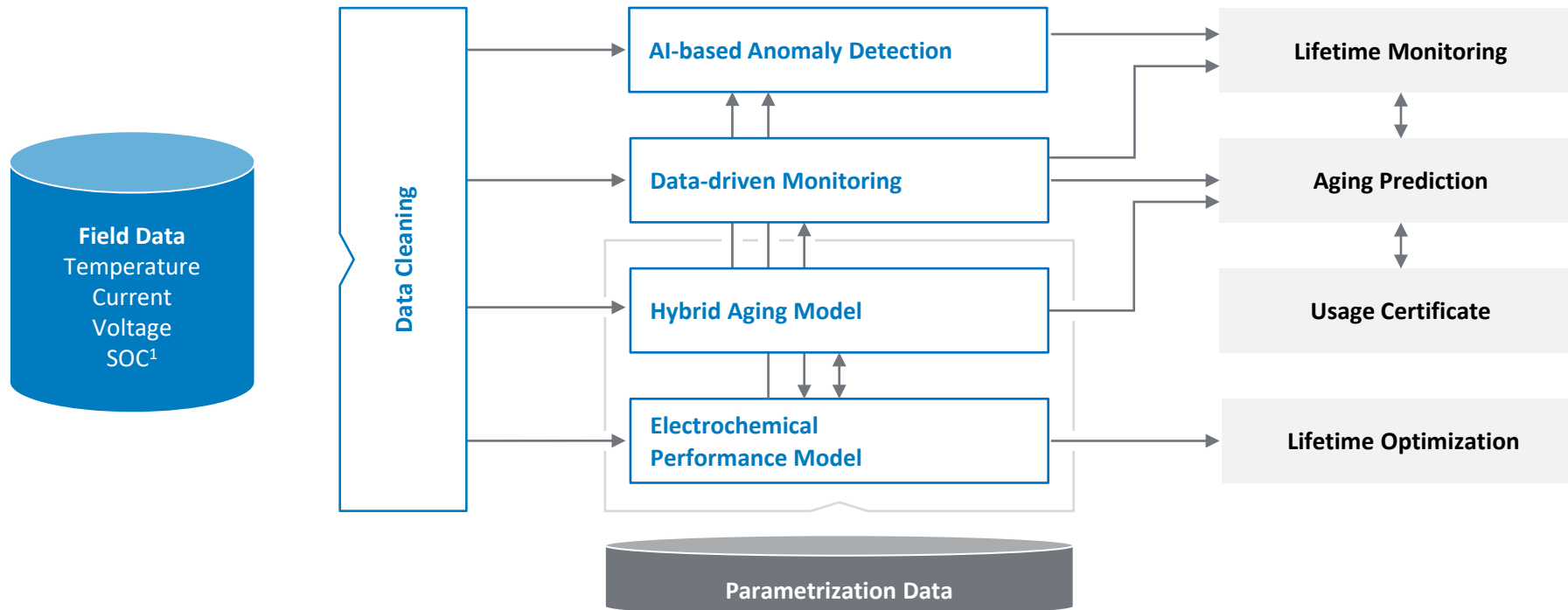


- ▶ Generate customer value by better understanding
- ▶ Improve performance or knowledge about SOH

# Predictive Maintenance for Components

## BitC - Solution combines Big Data, AI and Electrochemical Models

Optimize maintenance - reduce repair visit  
Determine components individual RuLc



### Battery / Algorithm / Cloud expertise

series availability ww including CN | market proven |  
> 180 patents applied for Battery in the Cloud only

### Bosch AI & system competence

Worldwide organizations for system integration and  
dedicated AI R&D organization | >> 800 AI patents

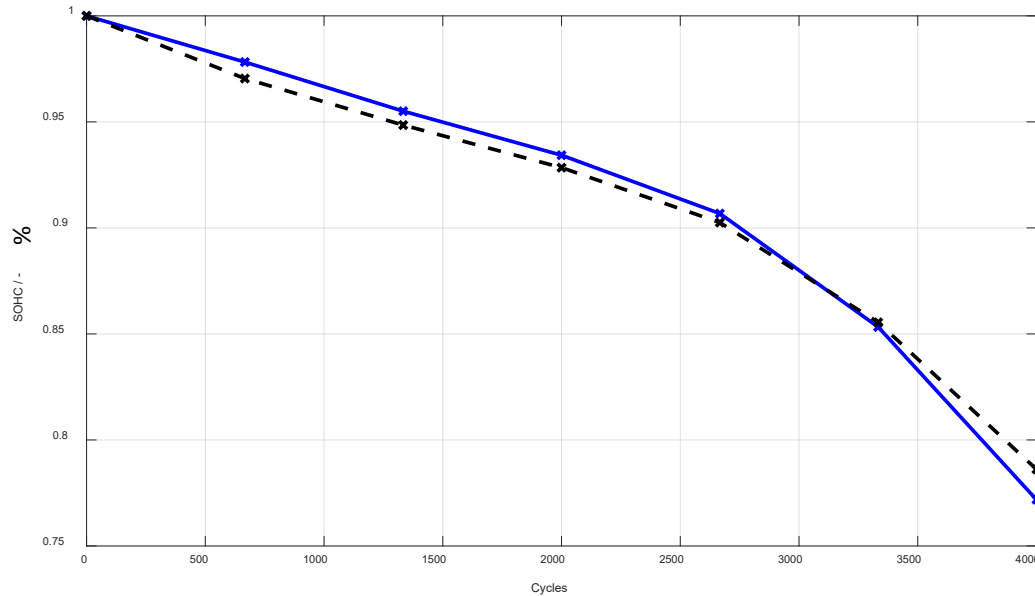
### Bosch electro-chemical competence

Broad library of material properties | Aging and  
performance model | Test infrastructure, labs

# Predictive Maintenance for Components

## BitC - Aging Prediction

Optimize maintenance - reduce repair visit  
Determine components individual RuLc



SOHC measurement

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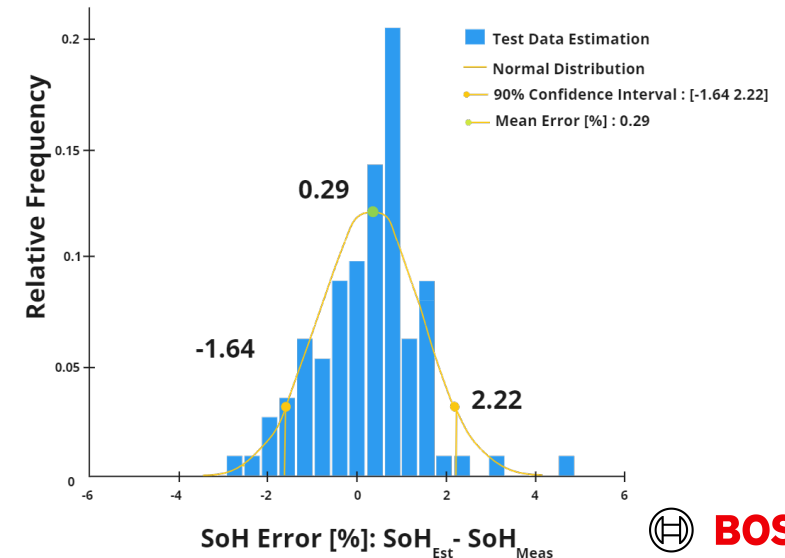
SOHC prediction

—

\*in case of availability of historic data

### Precise aging prediction

- Prediction of battery aging and remaining useful life (SoHc, SoHr) over years leveraging electrochemical and historic operations data\*
- Accurate results based on fusion of electrochemical and AI model
- Low residuum enables advanced planning of actions on field vehicles
- Due to advanced AI methods number of vehicles/battery data necessary to perform prediction is comparably low to extensive aging tests

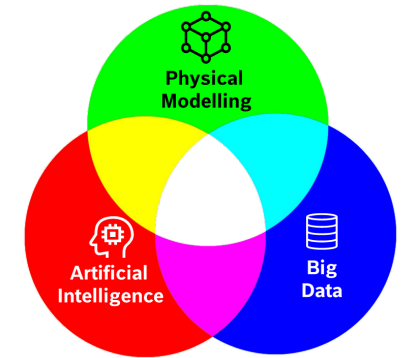


# Predictive Maintenance for Components

## Summary

### Predictive Maintenance

- Two use-cases avoid sudden breakdown and optimize maintenance process
- Predictive diagnostics based on components and system behavior
- Key element of prediction are hybrid model – physical and AI models
  - Physical understanding and modeling are base of the predictive functions
  - Application and optimization of the prediction with AI methods



### Predictive diagnostics for batteries

- Predictive Maintenance for starter battery and high voltage battery available
- Cloud based physical and hybrid models supported with edge pre-processing in the vehicle
- Common framework for interaction with customer and user