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# Save Energy with Splunk

Leverage Process and Energy Data  
to Optimize Industrial Processes

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# Your 3 Key Take Aways



## 1 COLLECT

heterogeneous sensor data  
from industrial processes in  
one data platform



## 2 ENRICH

and correlate sensor data  
with additional data sources  
to create meaningful context



## 3 ANALYZE

various data sources to  
optimize processes and  
increase efficiency

# About us

## Philipp Drieger (Splunk)

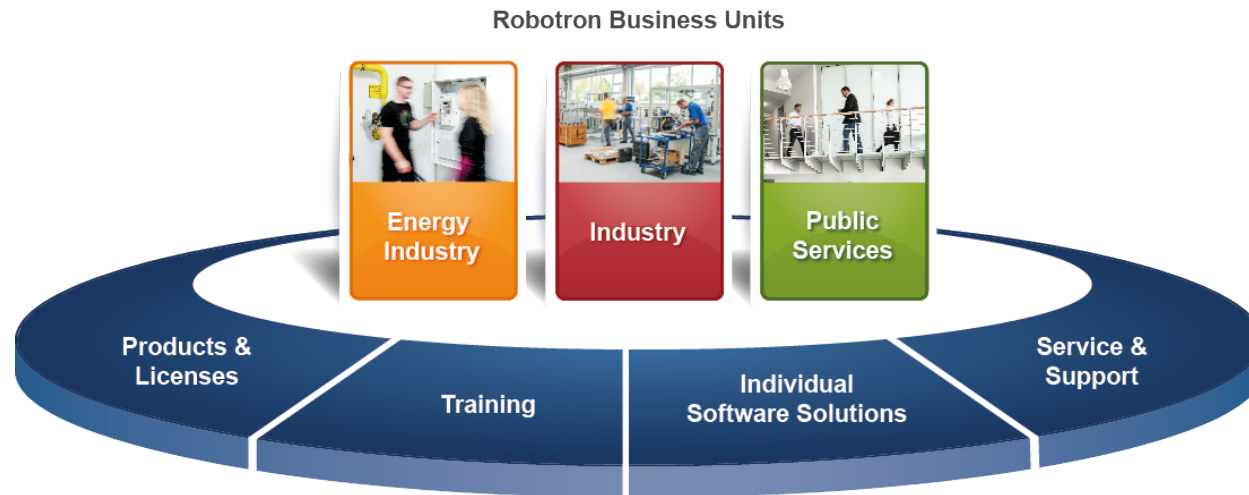
- Sales Engineer at Splunk
- Background in data visualization, analytics and 3D software development
- Experience in various industry verticals such as automotive, transportation and software industries.
- Proven fast time to value with Splunk winning Deutsche Bahn hackathon

## Matthias Ilgen (Robotron)

- Project manager and pre-sales engineer for business analytics
- Background in the area of information retrieval, text - and data mining
- Experience in various industry verticals such as life-science healthcare, manufacturing and automotive.
- Implementation of complex IoT solutions based on Splunk



# Facts about Robotron



- Methodical and technological responsibility
- Comprehensive expertise of industry-specific business processes
- Number of employees (Robotron group): 450

# Agenda

- Splunk as a data platform for industrial sensor data
- Bridging the gap: Combine energy and process data
- Use Case #1: Energy efficiency monitoring and optimization
- Use Case #2: Condition monitoring and predictive maintenance
- Conclusion & Outlook
- Q&A



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# Splunk as platform for industrial and IoT data

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# Splunk a World of Interconnected Assets



Transportation | Energy | Utilities | Building Management



Oil and Gas | Manufacturing

Sensors, Pumps, GPS, Valves, Vats, Conveyors,  
Pipelines, Drills, Transformers, RTUs, PLCs,  
HMIs, Lighting, HVAC, Traffic  
Management, Turbines,  
Windmills, Generators,  
Fuel Cells,  
UPS



Industrial Data



Retail | Home | Consumer



Telemedicine | Connected Cars

Wearables, Home Appliances, Consumer  
Electronics, Gaming Systems, Personal  
Security, Set-Top Boxes, Vending  
Machines, Mobile Point of  
Sale, ATMs,  
Personal Vehicles



Internet of Things

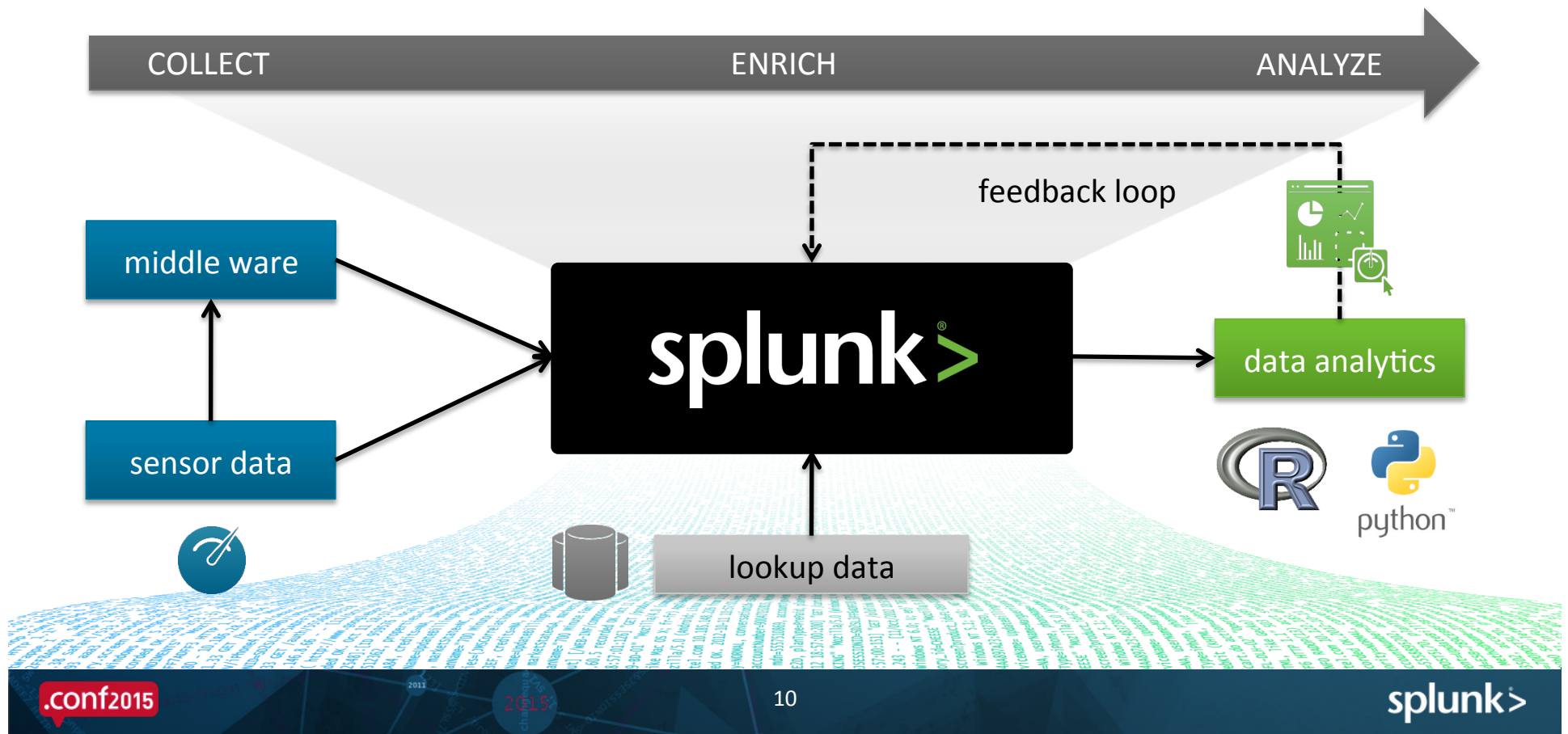
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# Splunk for 360 degree data view



# Typical Workflow for Analyzing Sensor Data



# 3 Common Ways to Analyze Sensor Data with Splunk

## 1 SPL

use out of the box SPL search commands to analyze your data

|                     |  |
|---------------------|--|
| stats               | calculate various statistics                         |
| (time)chart         | chart (time-series) events for viz                   |
| anomalies / outlier | detect unusual / outlier events                      |
| cluster / kmeans    | cluster events based on similarity / given cluster # |
| associate / arules  | identify correlations / relationships between fields |
| autoregress         | calc autoregression (for moving average)             |
| correlate           | co-occurrence between fields                         |
| contingency         | calc relationship between variables                  |
| predict             | prediction for time-series data                      |

## 2 APPS

leverage Splunk Apps to quickly onboard data and gain insights



<https://splunkbase.splunk.com/>

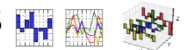
## 3 SCRIPT

create scripts or code with SDKs for advanced and customized solutions



NumPy  
Base  
N-dimensional  
array package

pandas



$y_{it} = \beta x_{it} + \mu_i + \epsilon_{it}$

# Cheat sheet: Splunk Commands for Analytics

| Splunk command      | What can I achieve with it?                          |
|---------------------|--|
| (stream)stats       | calculate various statistics                         |
| (time)chart         | chart (time-series) events for viz                   |
| anomalies / outlier | detect unusual / outlier events                      |
| cluster / kmeans    | cluster events based on similarity / given cluster # |
| associate / arules  | identify correlations / relationships between fields |
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Find out more: <http://docs.splunk.com/Documentation/Splunk/latest/SearchReference>



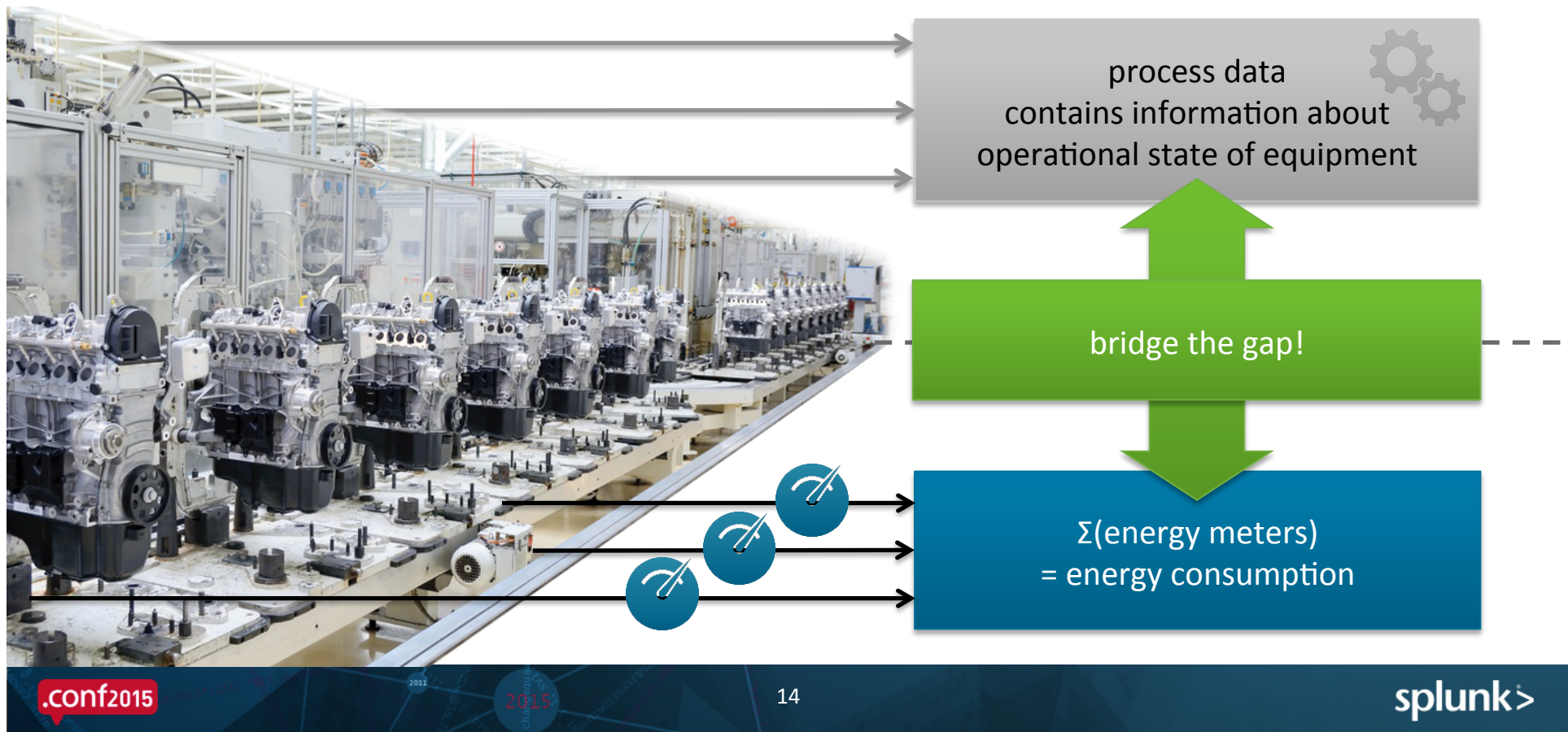


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# Energy Data & Process Data

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# Challenge: Optimize Energy Efficiency



# What is...

| Category  | Energy Data                   | Process Data              |
|-----------|-------------------------------|---------------------------|
| Time      | Equidistant time series       | Process event based       |
| Type      | Sensor data                   | Control data, sensor data |
| Semantics | Energy metrics                | Equipment behavior        |
| Source    | Energy logger, Equipment, EDM | SPS, SCADA, HMI, ...      |
| Format    | Variety of formats            | Variety of formats        |



## Energy Data

Energy consumption

86348 24.03.15 23:59:59 140808,297;  
140746,031;140919,500;



## Process Data

Production status

24-03-2015 01:00:59 EPIP02-03-A;SB,PPR PR;  
PRODUCTION;PR;aRTC: accounted transaction  
(equip02\_evnt\_job\_unit01);;;;;;;;;;0,014;753,000

correlation over time (join)

Use cases

- Transparency of equipment on shop floor level
- Discover process weaknesses
- Condition based and predictive maintenance
- Optimization of energy efficiency of equipment
- Optimization of energy purchasing process (forecast / predictive)
- ... etc ....

Increased efficiency  
Saved energy  
Saved \$\$\$





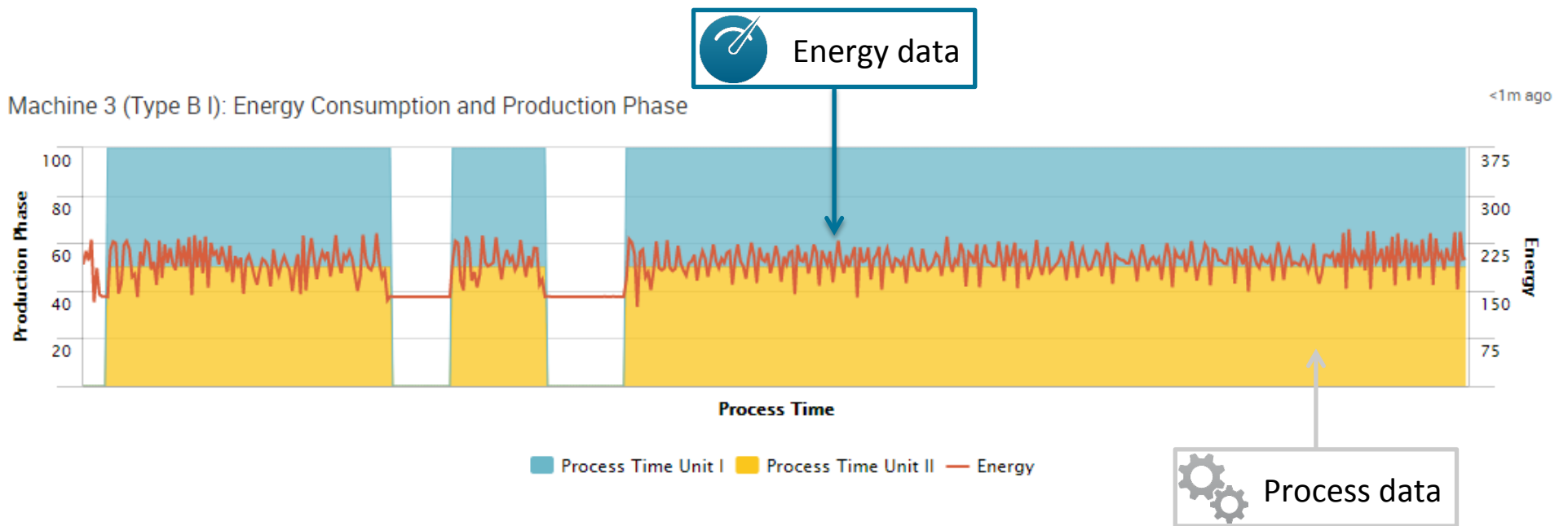
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# Energy Efficiency Monitoring & Optimization

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# EEE Monitoring & Optimization

Correlating energy and process data



# Energy Efficiency Monitoring

- Optimization of energy efficiency for production
- Reduction of energy consumption of non-value-adding activities
- Optimization of production schedule of similar equipment
- Reduction of specific energy consumption per produced item

$$\text{Energy Efficiency of Equipment (EEE)} = \frac{\Sigma(\text{value-added energy consumption})}{\Sigma(\text{total energy consumption})}$$

# Energy Efficiency of Equipment (EEE)

High Level Overview: Finding efficiency issues at a glance

All Machines and Units



**Machine 1 (Type A I):**

Energy Efficiency:

**0.403**

Total Energy (kVAh):

**3631.77**



**Machine 2 (Type A II):**

Energy Efficiency:

**0.62**

Total Energy (kVAh):

**3250.49**



**Machine 3 (Type B I):**

Energy Efficiency:

**0.842**

Total Energy (kVAh):

**4276.82**



**Machine 4 (Type B II):**

Energy Efficiency:

**0.812**

Total Energy (kVAh):

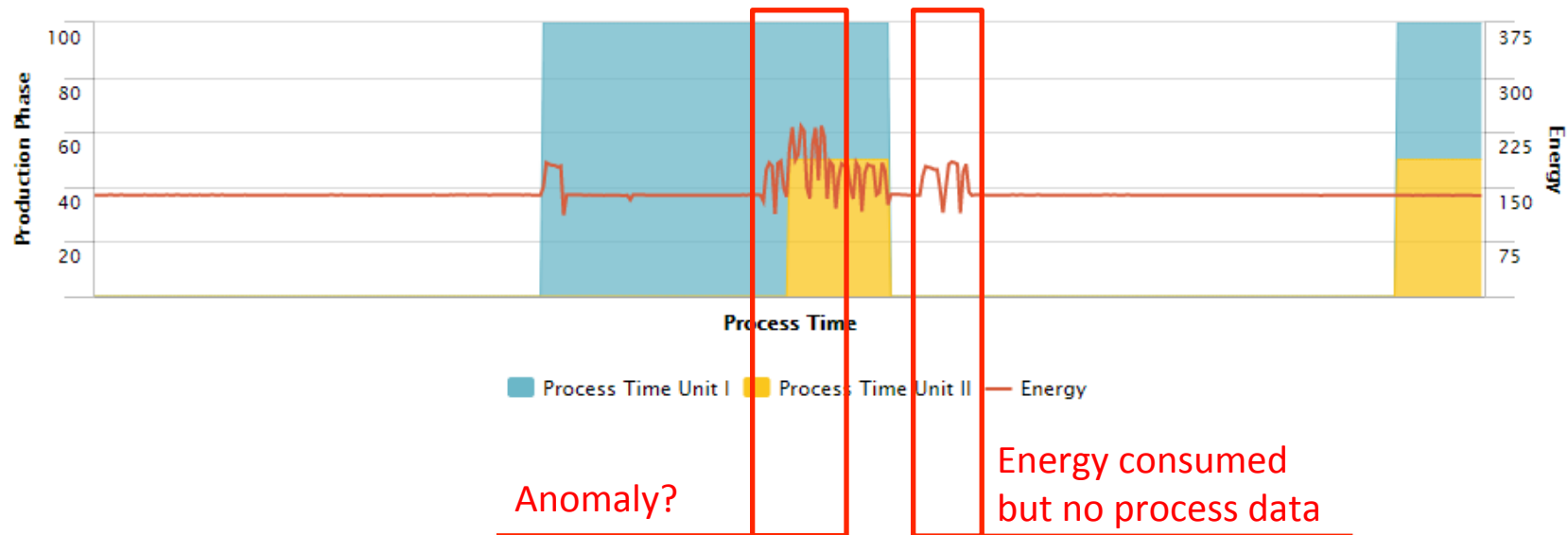
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# EEE Monitoring & Optimization

Detect process weaknesses: identifying anomalous patterns

Machine 1 (Type A I): Energy Consumption and Production Phase



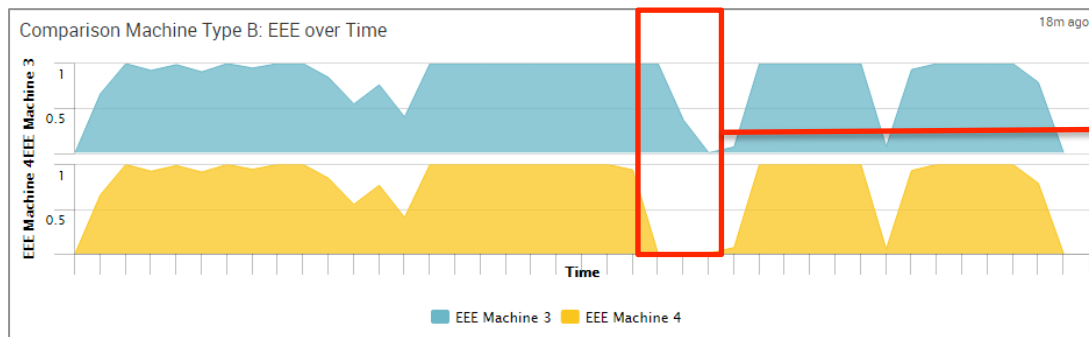
# EEE Monitoring & Optimization

Detect process weaknesses: optimize stand-by times

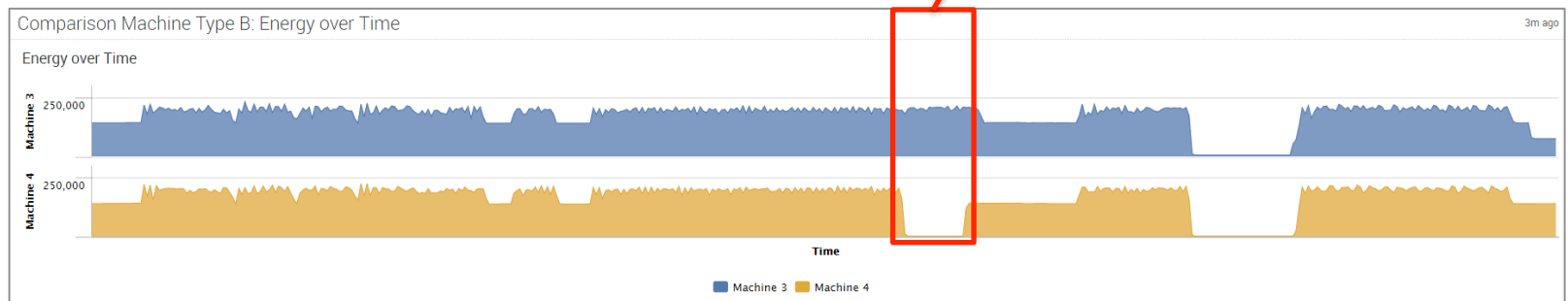


# EEE Monitoring & Optimization

Energy consumption over time: benchmark different equipment of same type



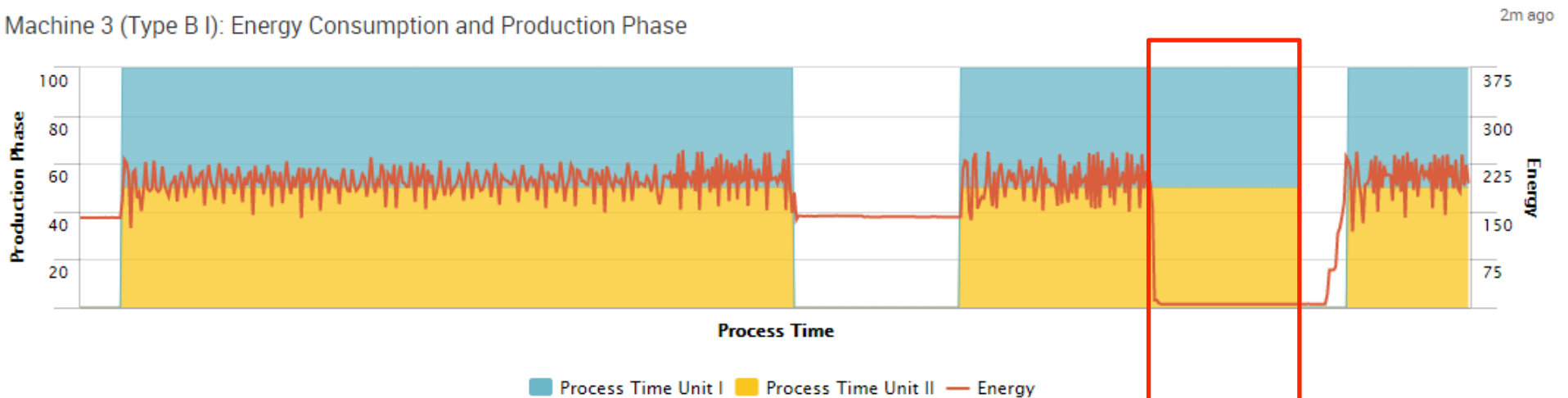
Visual correlation:  
Detect differences!



# EEE Monitoring & Optimization

Further use cases: “findings by accident”

Machine 3 (Type B I): Energy Consumption and Production Phase

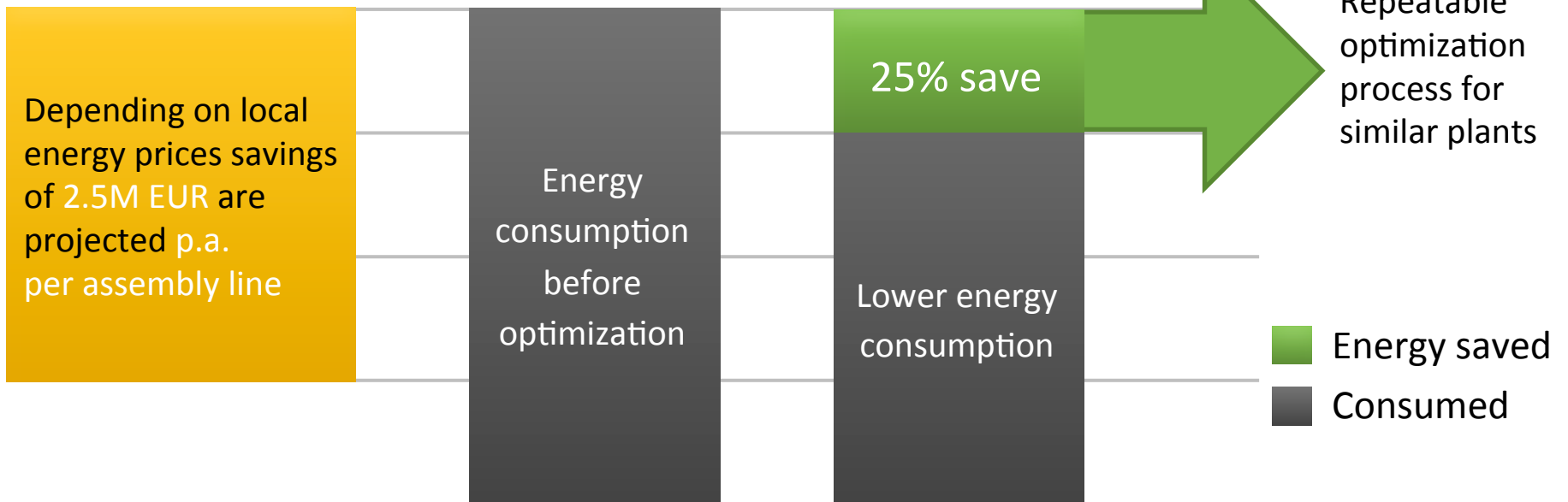


Find data quality issues:  
reported false process status



# Energy Savings

Up to 25% per facility





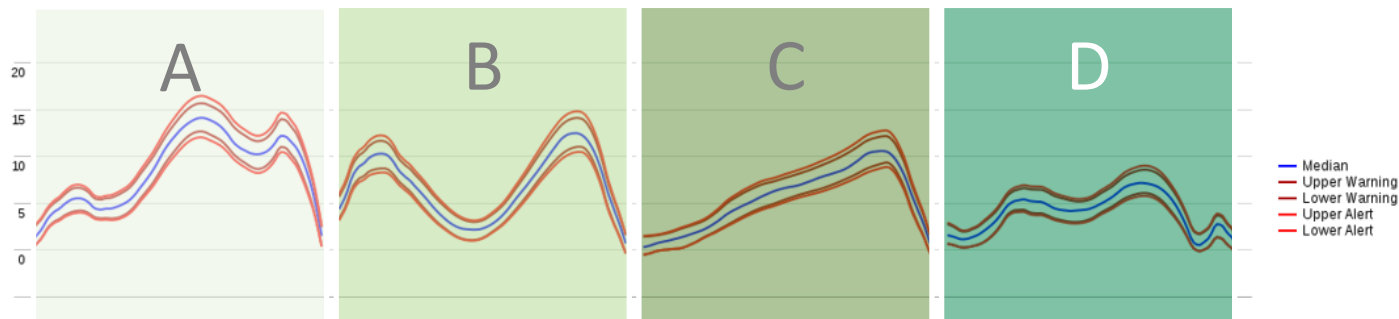
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# Condition Monitoring & Predictive Maintenance

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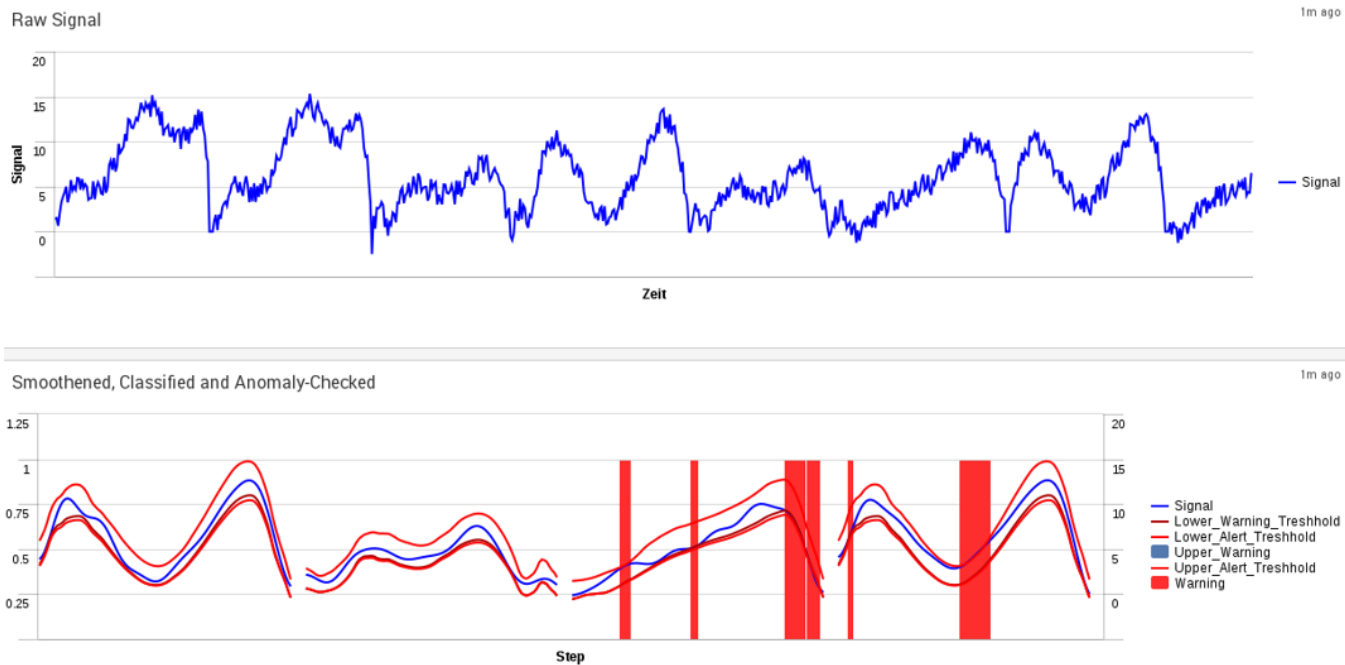
# Energy and process data for maintenance

- Energy not just a optimization target – but also an influencing factor for maintenance scenarios (rapid impact factor)
- Map low level process status to particular energy consumption profiles and learn normal states and boundaries from raw signal



# Condition Monitoring & Alerting

## Anomaly detection and proactive monitoring



# Predictive Maintenance

Predict anomalies for a particular process step

Predict process steps  
In which an error might  
happen

Cycles Until Next Alert

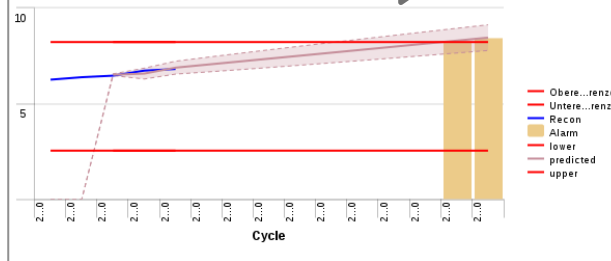
1m ago

! 9

Extrapolation of process  
steps with integrated  
predict function or  
other regressions models

Extrapolation Next Alert

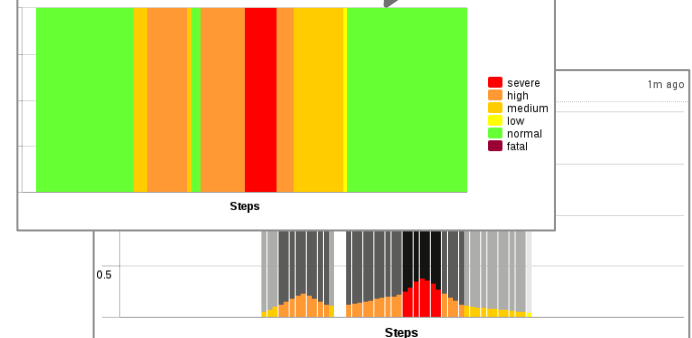
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Heatmap shows  
(recommend) time span  
in which an error might  
happen

Urgency of Maintenance - Heatmap

1m ago







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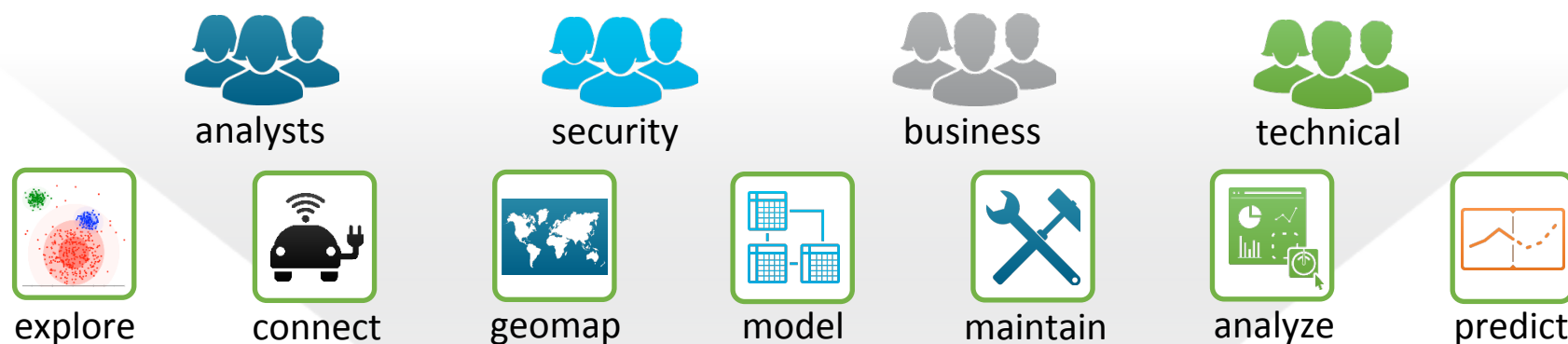
# Outcome and Outlook

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# Summary & Outlook

- Generic and equipment-independent approach
- No data transformation and model mapping in advance
- Applicable for “old” equipment (without particular sensor installation)
- Out-of-the-Box Splunk data models for energy and process data
- 360° view - several kinds of visualizations
- Own Splunk commands for numeric operations and machine learning
- Enhanced time series forecasting for optimization of energy purchasing

# Robotron Architecture for Industrial Data Analysis



Common Information Model for IoT purposes

**splunk>enterprise**

Robotron Switching Server

Raw sensor data

lookup, process data  
and other machine data

data analysis  
using SPL, R  
and Python



Databases Applications Servers Networks

python™



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Meet me @  
IoT Pavillon!

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THANK YOU

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