

Big Data for Operator Support in Chemical Plants

Introduction







### Chemial Industry – A case for Big data





- > 300 GB measurement data p.a. in a single refinery
- 400 GB alarms & events p.a. in a single petrochemical plant

# ■ High Velocity, e.g.:

■ 66.000 sensor with sampling rates between 1s – 60s

# High Variety, e.g:

Time-Series, Log files, unstructured text, video data

# ■ Low Veracity, e.g:

time-synchronisation, faulty measurement, missing data

















### Chemical Industry – A challenge for Big Data



- Challenging problems für data analytics more machine learning then simple statistics
- Data collection processes not optimized for Big Data Analytics
- High efforts for data exploration due to data silos with unstructured and inconsistent references
- High efforts for data-preparation and cleansing due to interrelations unknown to the data analyst









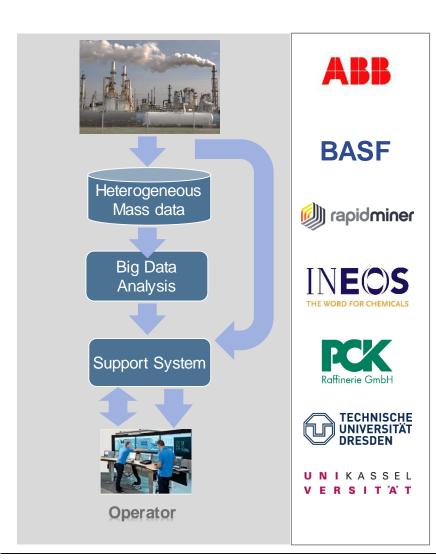












### **Objective: Operator Support functions**

- Early Warnings
- Ad-hoc Analysis
- Decision Support

### Approach: Integrated Analysis of all plant data

Measurements, engineering data, electronic shift books,...

### **Research Topics**

- Algorithm development
- Indexing of and search in process data
- Integration into real-time plant operation
- Big data technologies and architecture
- User Centered interaction concepts









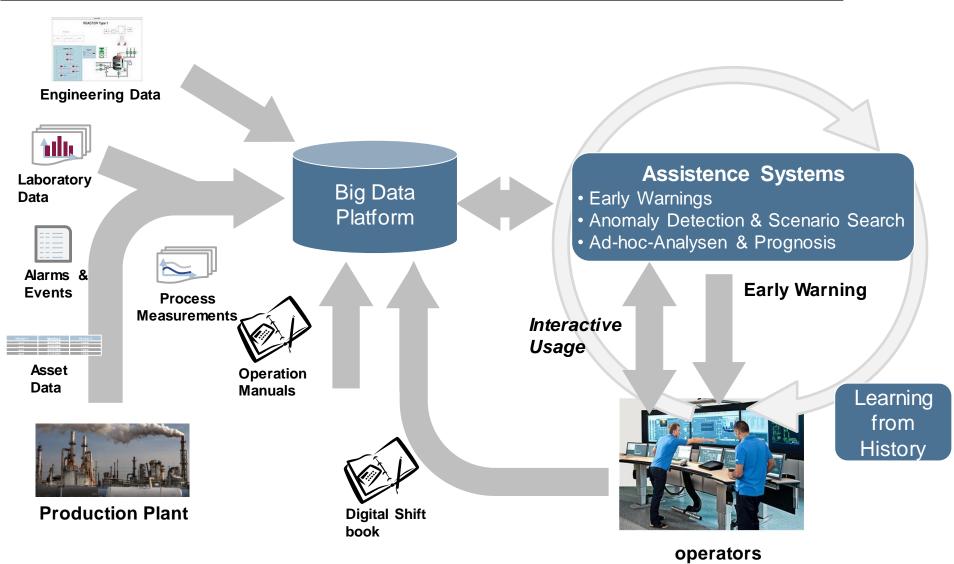








# **FEE – Data and System Landscape**











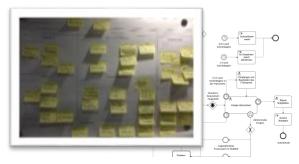






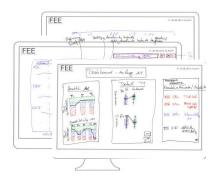






#### 1. Scenario Identification





2. Paper Prototypes



4. PoC Data Analytics



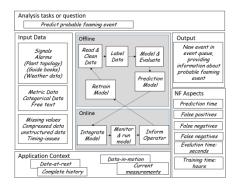






5. Refined Mock-ups





# 3. Analysis Workflows & Non-Functional Requirements

















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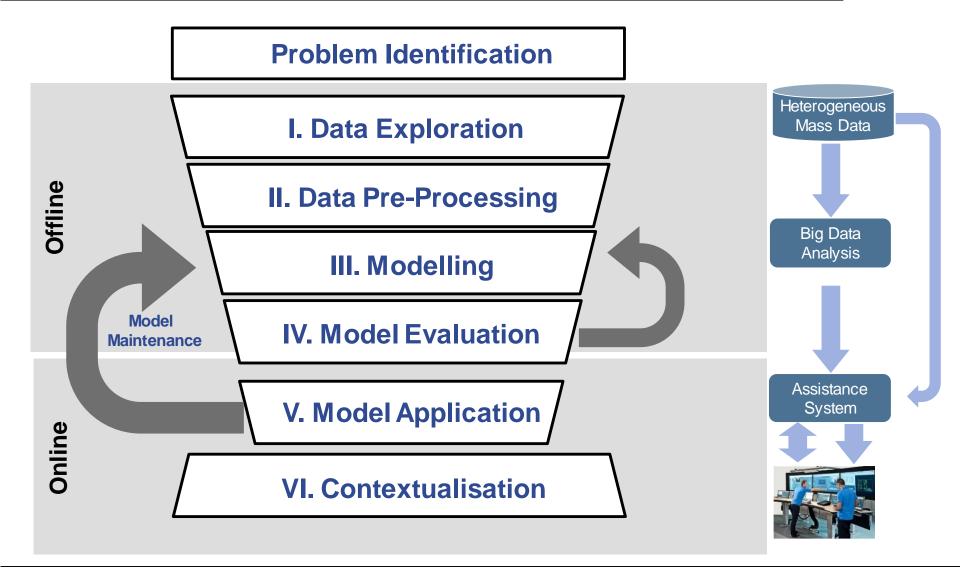
Szenario – From Big Data to Smart Data







### **Life Cylce of Operator Support Function**

















#### **Current State:**

#### Who:

Operators in control room and in the field

#### What:

Foaming in a process column results in increase pressure and risk of spillover. Anti-foaming agent needs to be added manually.

#### How:

Monitoring relevant signals in the control room

#### **Problems:**

- (1) Risk of not recognizing foaming early enough
- (2) Foaming is a fast process actions are always taken under time pressure
- (3) Unexperienced operators might not recognize the situation or do not know how to react

#### **Desired State:**

#### **FEE Support:**

Early information about certain or probably foaming in the new future.

#### Desire:

- (1) Timely information latest 30 minutes before the foaming
- (2) Clear and specific instructions, no need for diagnostics activities
- (3) High prediction rate, few false alarms







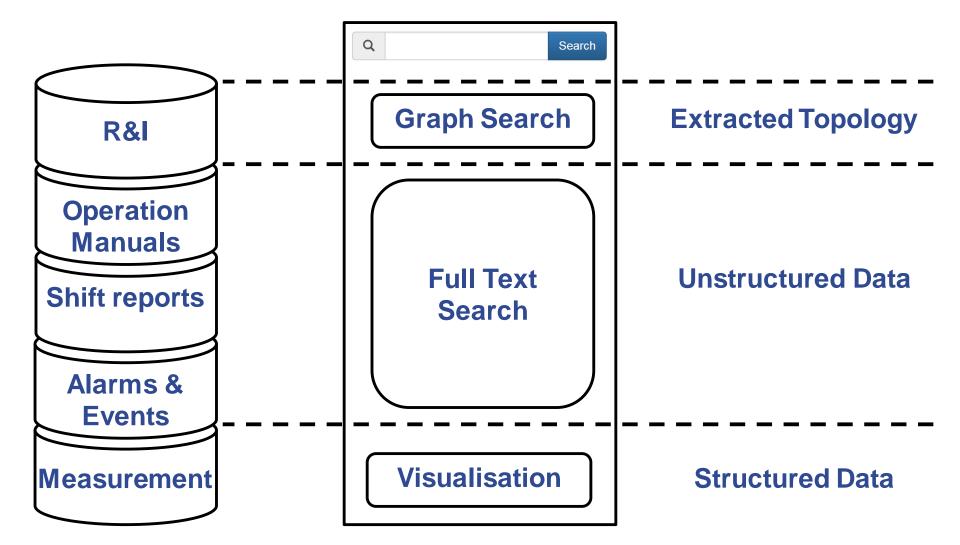


















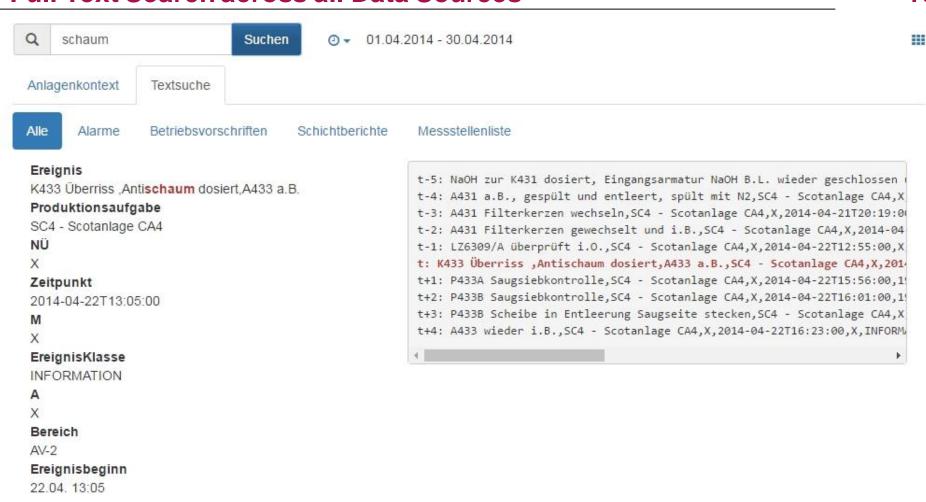








### Full Text Search across all Data Sources



### Simple access to data by full text search









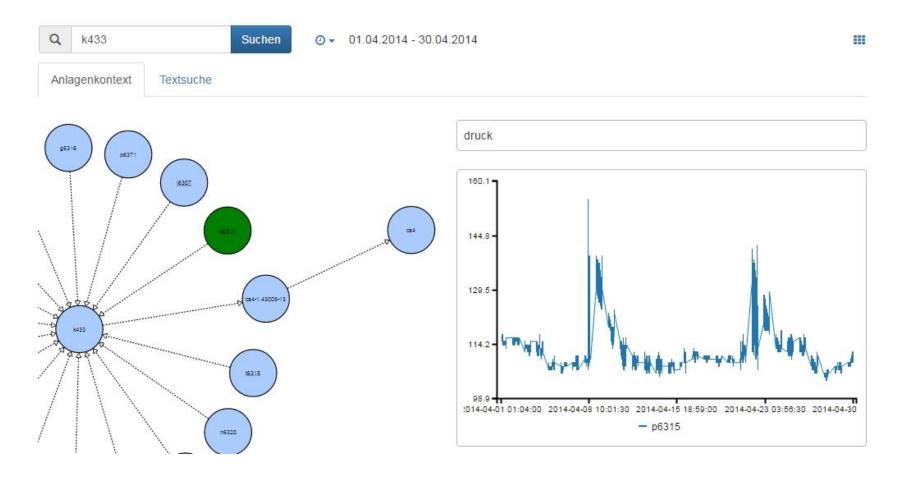












Graphical Exploration of data based on derived plant topologies









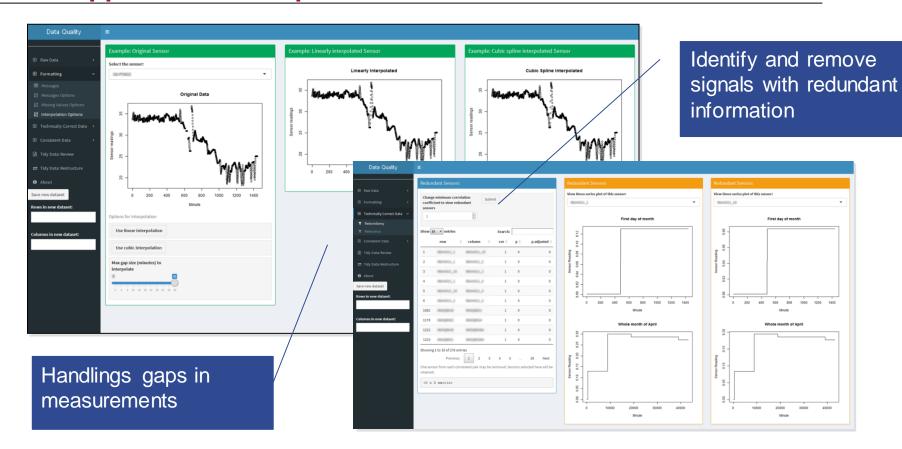








### **Tool supported Data Exploration**



# **Speed-up typical data cleansing & selection tasks**





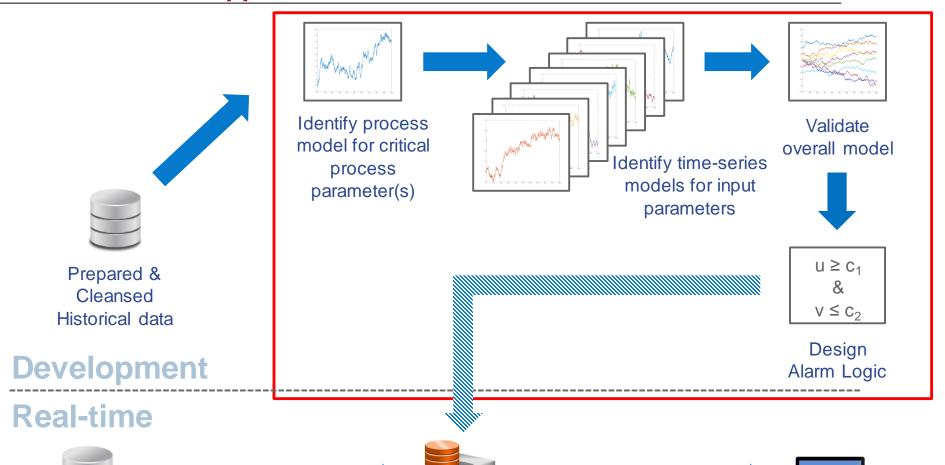














Cleansed

Online data









**Predictive Alarming** 



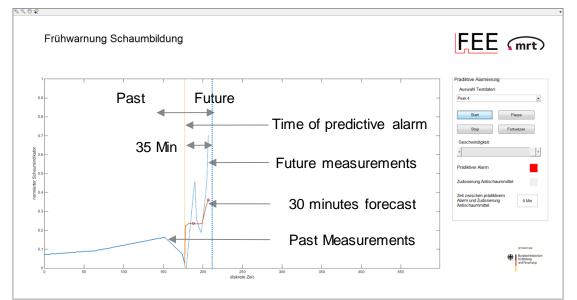


**FEE Operator Screen** 

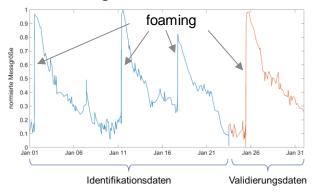


### Case study: foaming detection in SCOT plant

- Automated selection of significant input signals and model terms for ARX process model
- Automated selection of significant model terms for AR time-series models
- Overall validation by iterative multi-step prediction
- Simple alarm logic on predicted signals (threshold for signal amplitude and signal gradient)
- Predictive Alarming from Engineering Perspective



#### Critical signal:



- Sampling: 1 Min
- Measurements per signals: 44641
- Potential Input Signals: 29
- Significant Input signals: 7
- Timeliness of predictive alarm: 35 minutes

















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Scenario – Anomaly Detection: Big Data for rare events





#### **Current State:**

#### Who:

Operator in the control room (and process engineers)

#### What:

Monitoring of the plant in ,calm' situations

#### How:

- Browsing operator screens and trend display for suspicious signals
- Is only done in ,calm' situation without stress

#### **Problems:**

- (1) Risk to simply overlook a suspicious signal
- (2) Monitoring without broad coverage in stressful situations
- (3) Difficult for unexperienced operators to judge the , suspiciousness' of signals

#### **Desired State:**

#### **FEE Support:**

Identify suspicious signals and providing relevant data for diagnosis

#### Desire:

- (1) Fast visual impression on abnormalities in the process
- (2) Put into context to historical ,normal' and ,abnormal' signal paths
- (3) Providing extended context (relevant alarms, operator notes, documents)





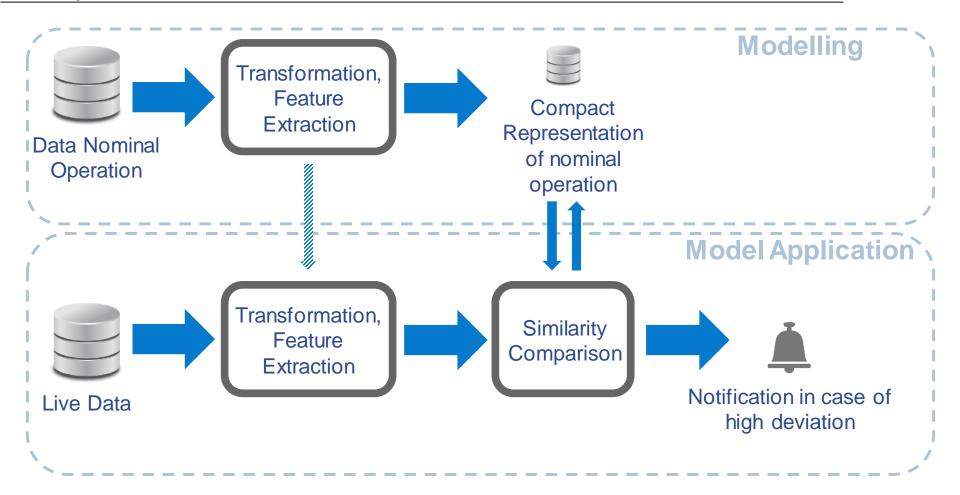






















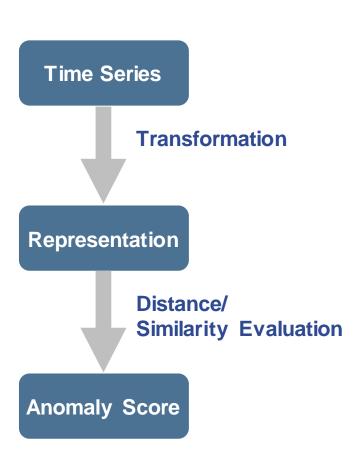


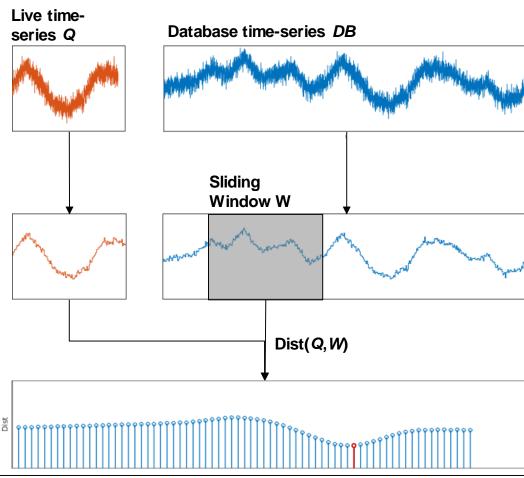




### **Subsequence Matching basierte Anomaliedetektion**

The distance between a live data time-series and the most similar subsequence from historical database is used to calculate the anomaly score.















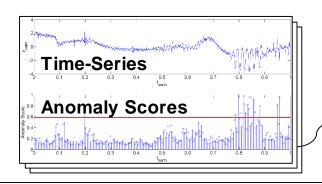




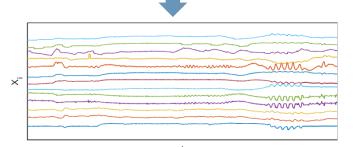


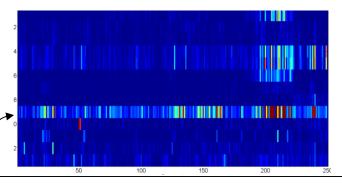
- Continuously operated butadiene plant
- One (known) singular anomaly
- **High Data Volume:** ~1000 measuring points with sampling rate of 1 minutes over two years
- **Heterogeneous:** Pressure, flows, levels, analyzer, temperatures, varying compression over time and different from time-series to time-series
- Nonstationary: Frequent load changes
- Data Selection:
  - Data selection without expert knowledge: Elimination of redundant and constant time-series to 104 measuring points
  - Data selection by expert knowledge: 13 measuring points (shown)

 Visualization of calculated anomaly scores in a heat map



























Information available in unstructured formats







 Objective: Support operator in finding information by suggestion of context-sensitive search terms

Antischaum Durchfluss
Desorber Kolonne
Pumpe 324 Kopfdruck









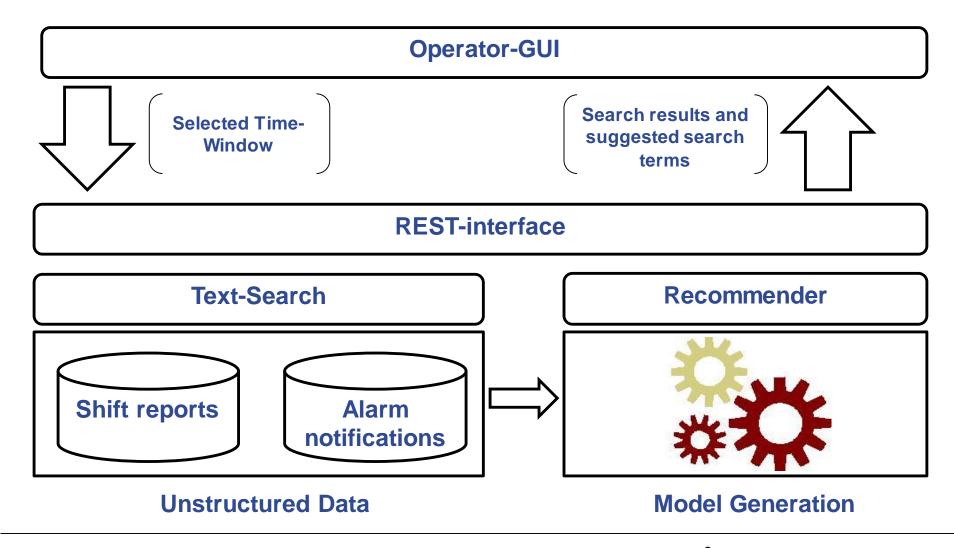




















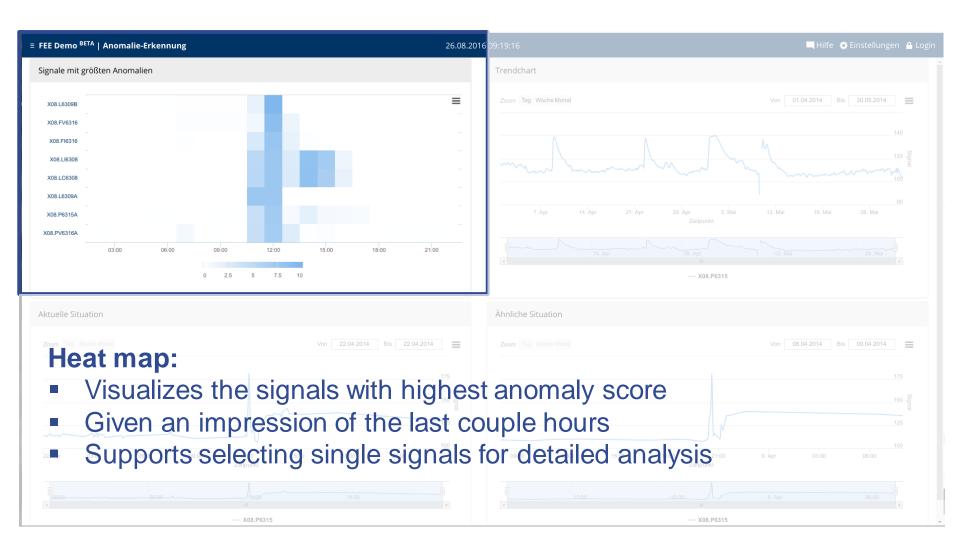








### **Operator Interface – Suspicous Signals (1)**



















# **Operator Interface – Suspicous Signals (2)**































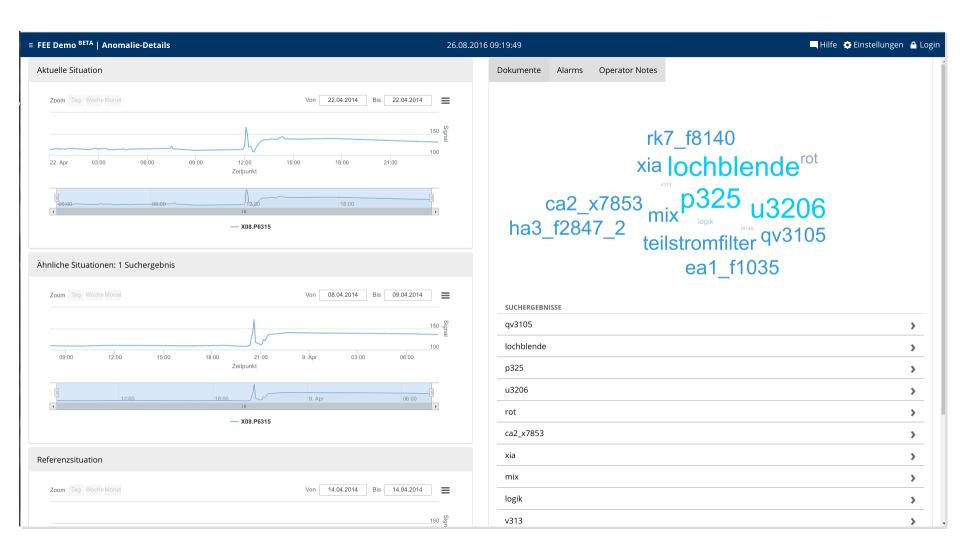








### **Operator Interface – Suspicous Signals (4)**











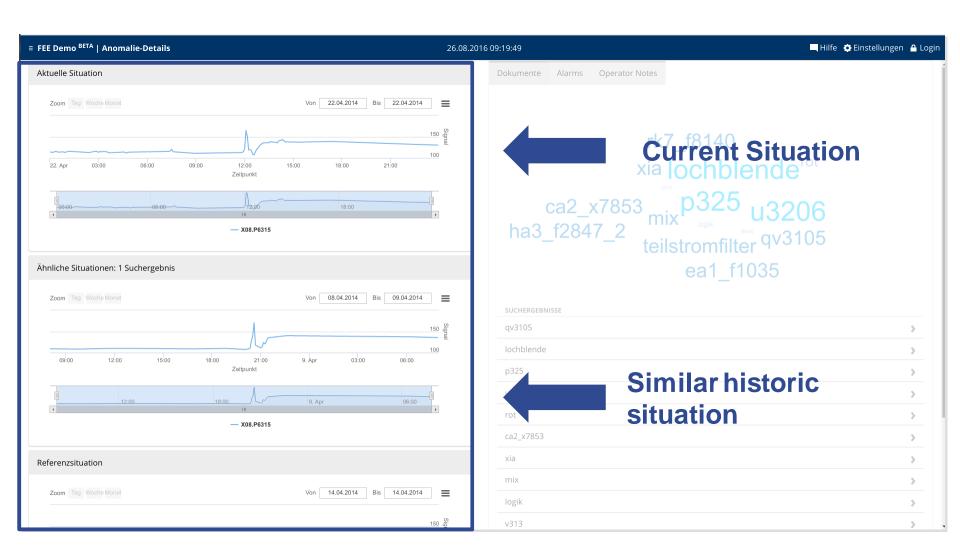








### **Operator Interface – Suspicous Signals (5)**









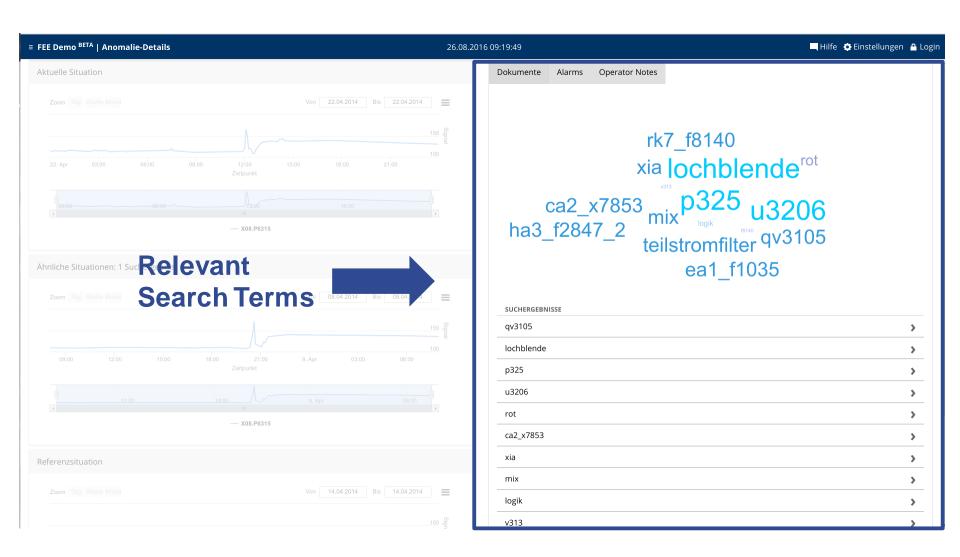




























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# **Summary**







- What has been shown
  - Transfer of (big) data analytics into the context of chemical industry
  - Challenges of a big data architecture for chemical plants
  - Solution approach with two typical scenarios (Event prediction and anomaly detection)
- Next steps
  - Work on additional application scenarios
  - Further refinement of methods and demonstrating transfer to other plants
  - Demonstration of functionality in the plant context













