

Big Data Analytics for Process and Quality Monitoring in Gas Metal Arc Welding of Automotive Parts

Joining Smart Technologies International Automotive Conference 08.05.2019

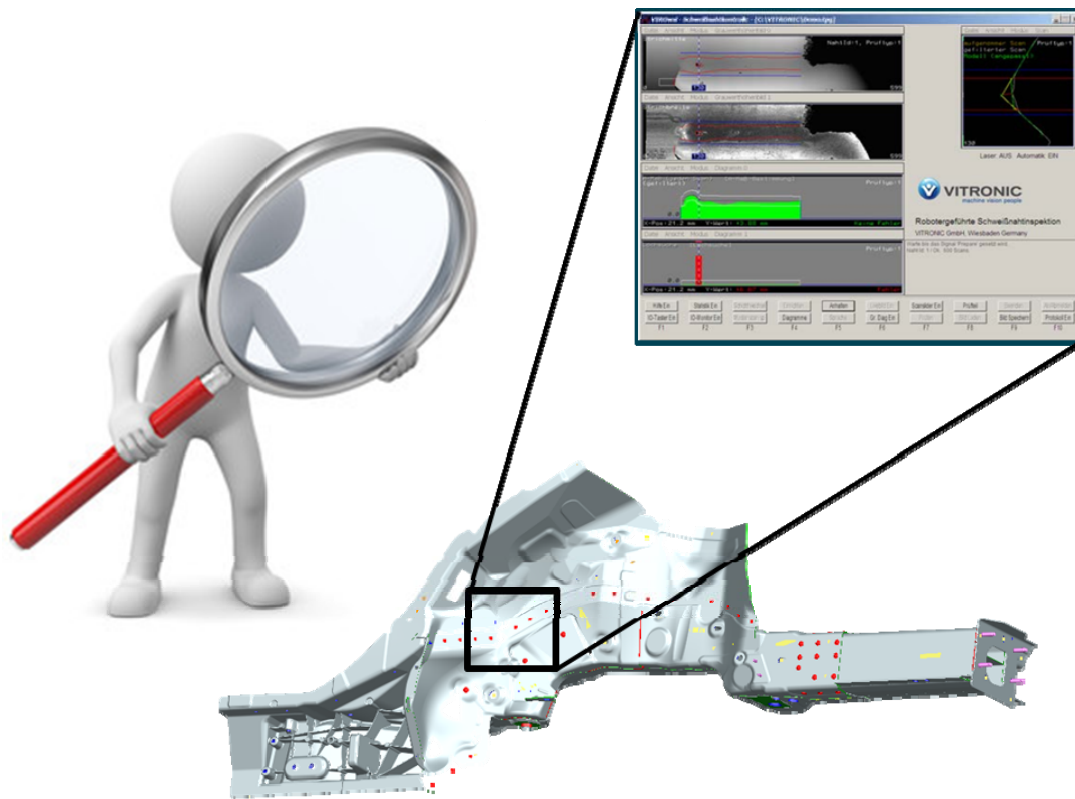
Luttmer, Michael; Thaler, Heiko; Dongus, Jürgen; Process Engineering

Mercedes-Benz

The best or nothing.

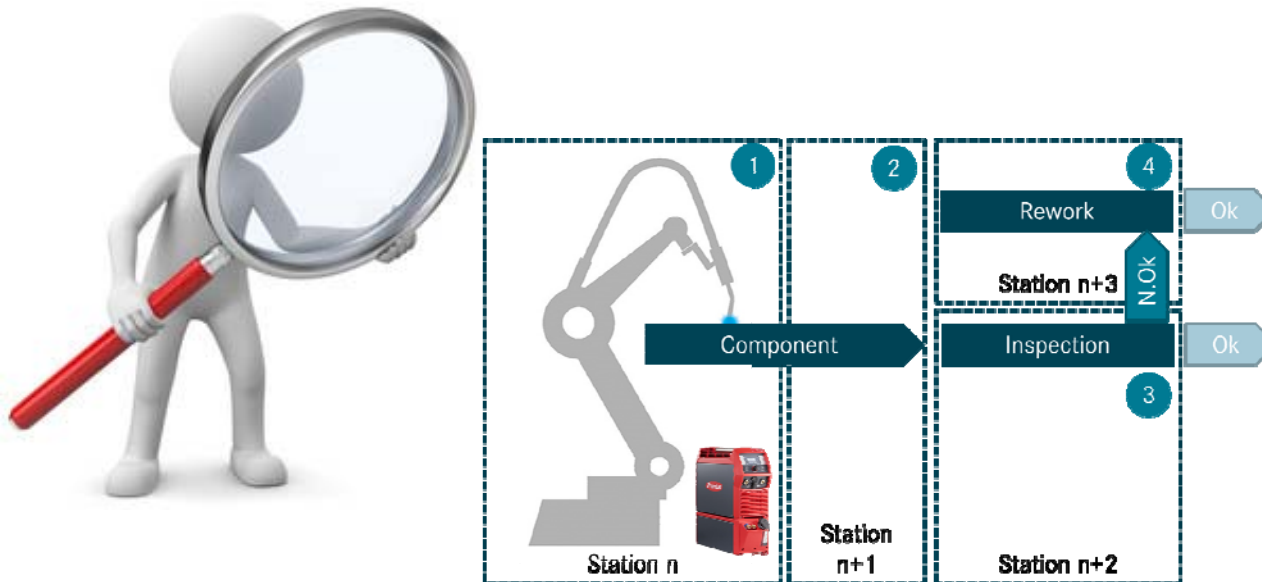


Fully automated optical weld seam inspection systems (e.g. VIRO VSI from VITRONIC)



- Only optical evaluation of the weld seam
- Measuring principle: numerically error based
- Additional inspection station
- No contribution to increase process capability

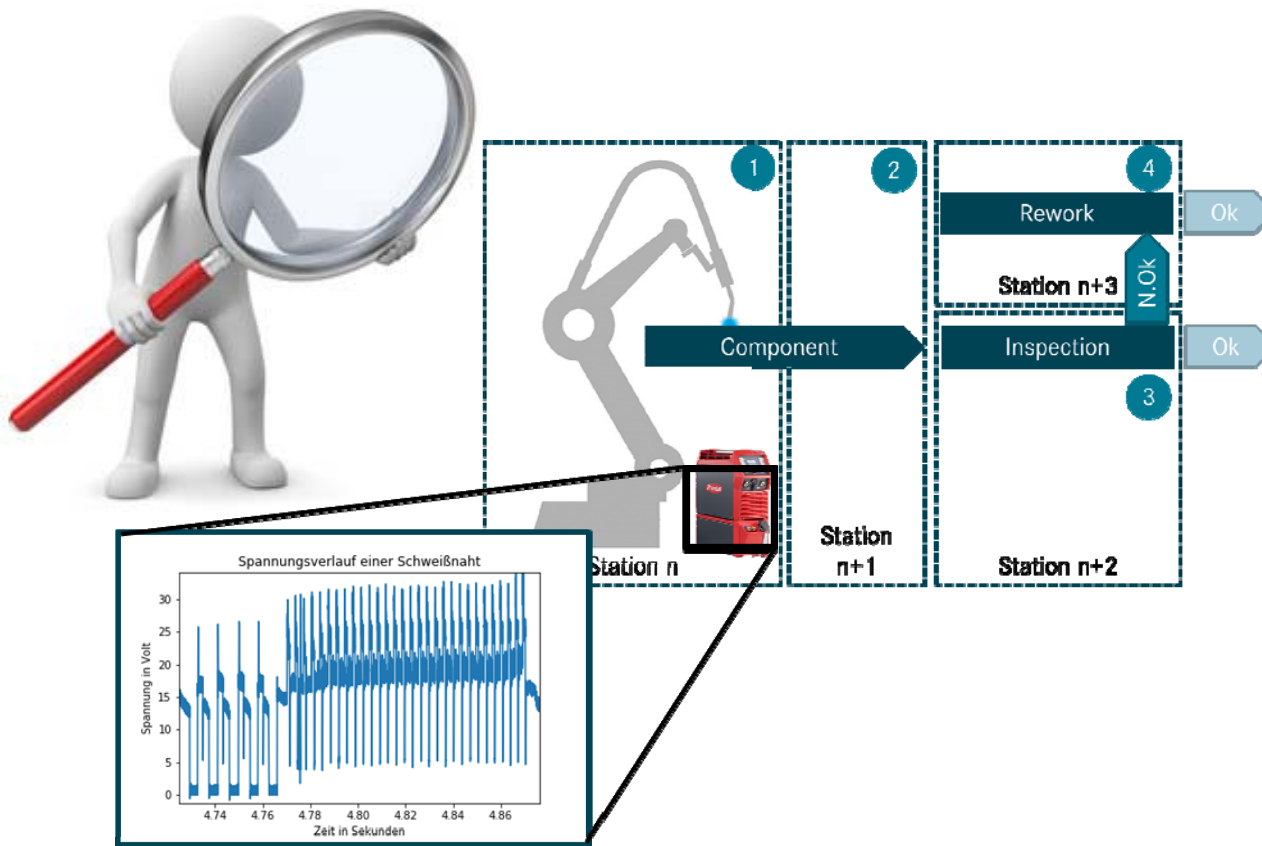
Fully automated optical weld seam inspection systems (e.g. VIRO VSI from VITRONIC)



- Only optical evaluation of the weld seam
- Measuring principle: numerically error based
- Additional inspection station
- No contribution to increase process capability

Data Analytics on process data to increase weld seam quality and process capability

Fully automated optical weld seam inspection trough systems such as the VIRO VSI from VITRONIC



- Only optical evaluation of the weld seam
- Errors based on measuring principles
- Additional inspection station
- No contribution to increase process capability

Data Analytics on process data to increase weld seam quality and process capability

Approach: Find features with significant correlation to your target



Consumption [ltr/ 100km]

10

15

Time [min]

60

45

Distance

Stuttgart - Karlsruhe

Target

Who was speeding?



Approach: Find features with an significant correlation to your target



Color

yellow

blue

Add-ons

climatronic

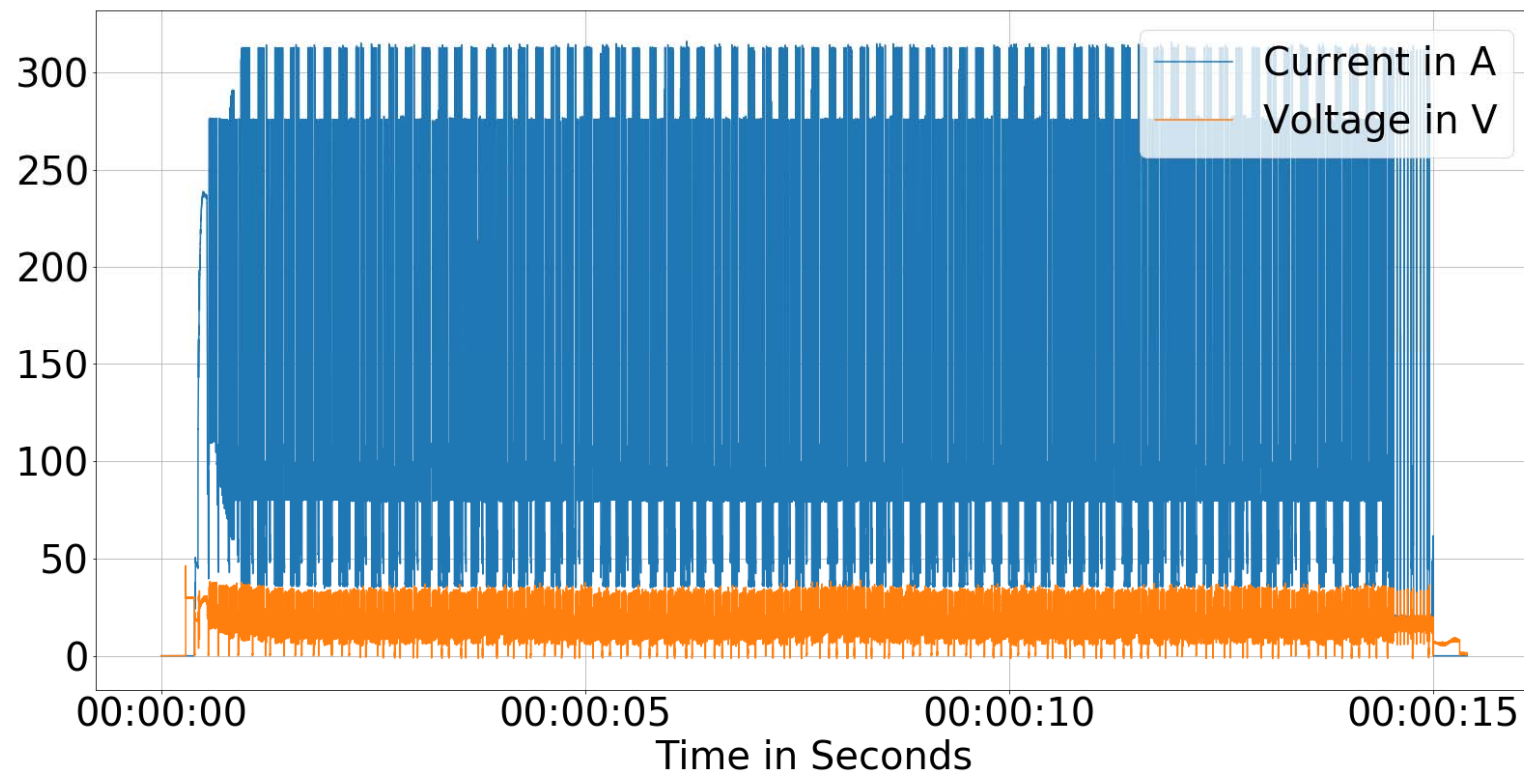
bending lights

Distance

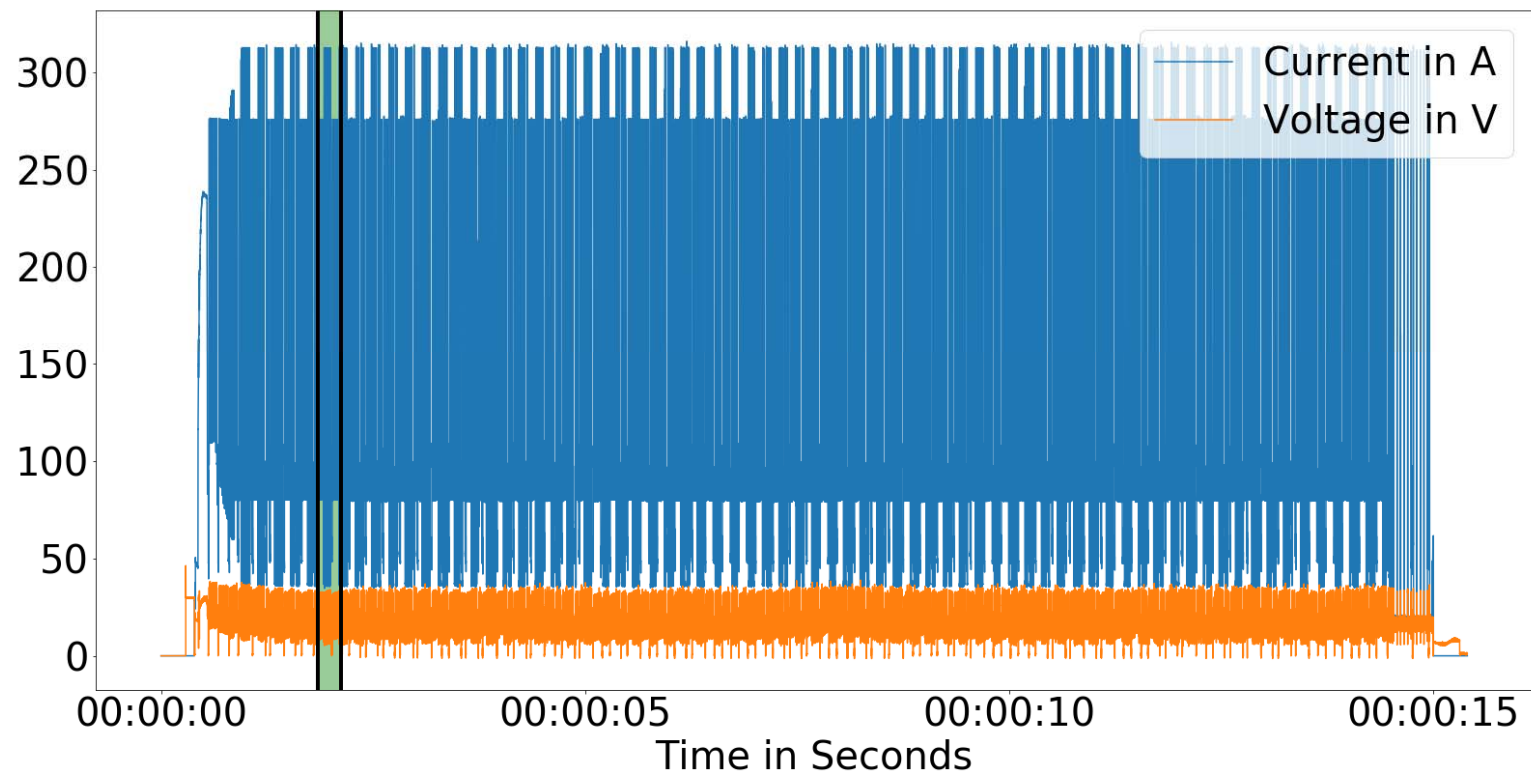
Stuttgart - Karlsruhe



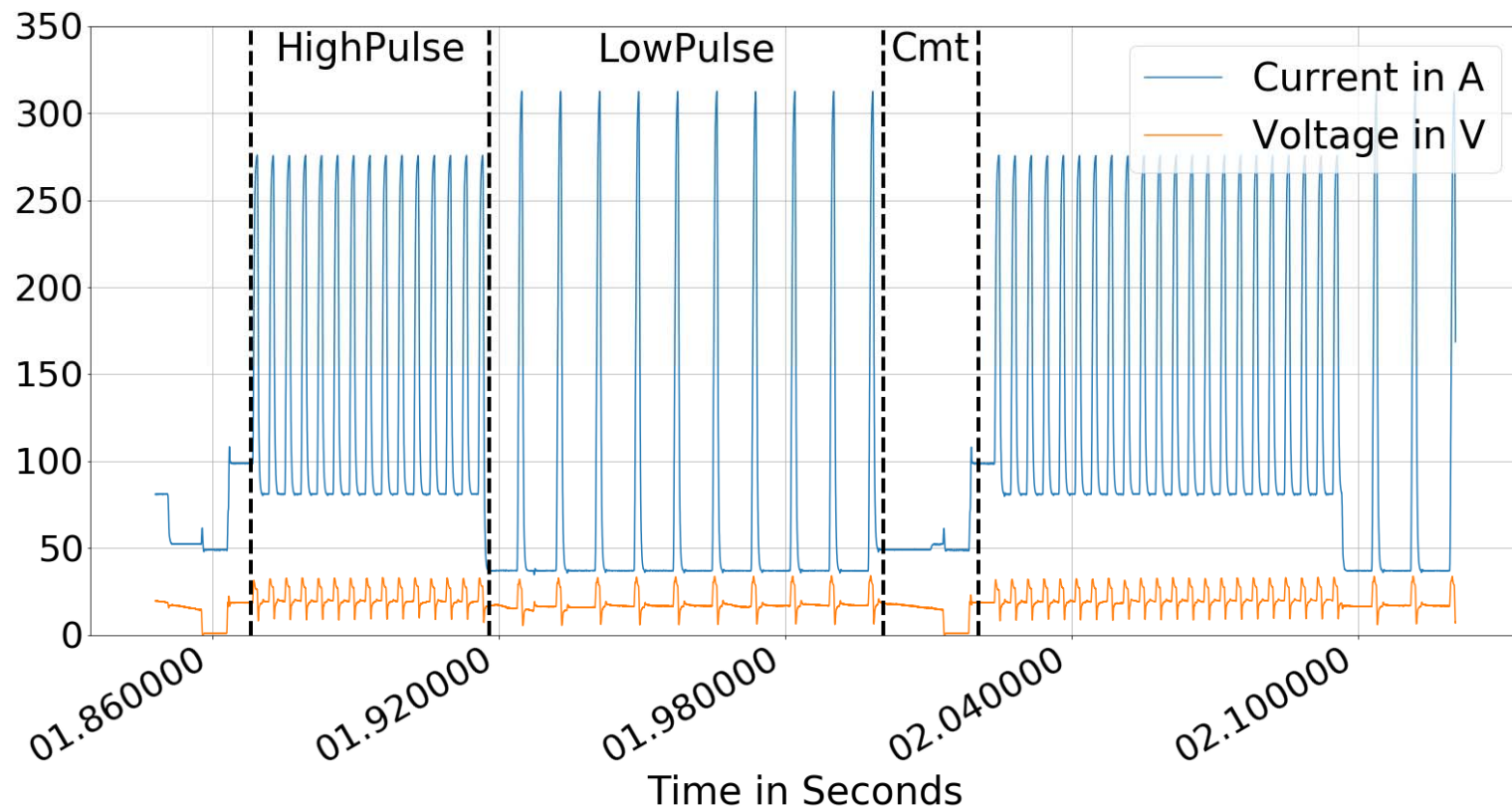
The „modern“ gas metal arc welding process is a mixture of different process types



The „modern“ gas metal arc welding process is a mixture of different process types

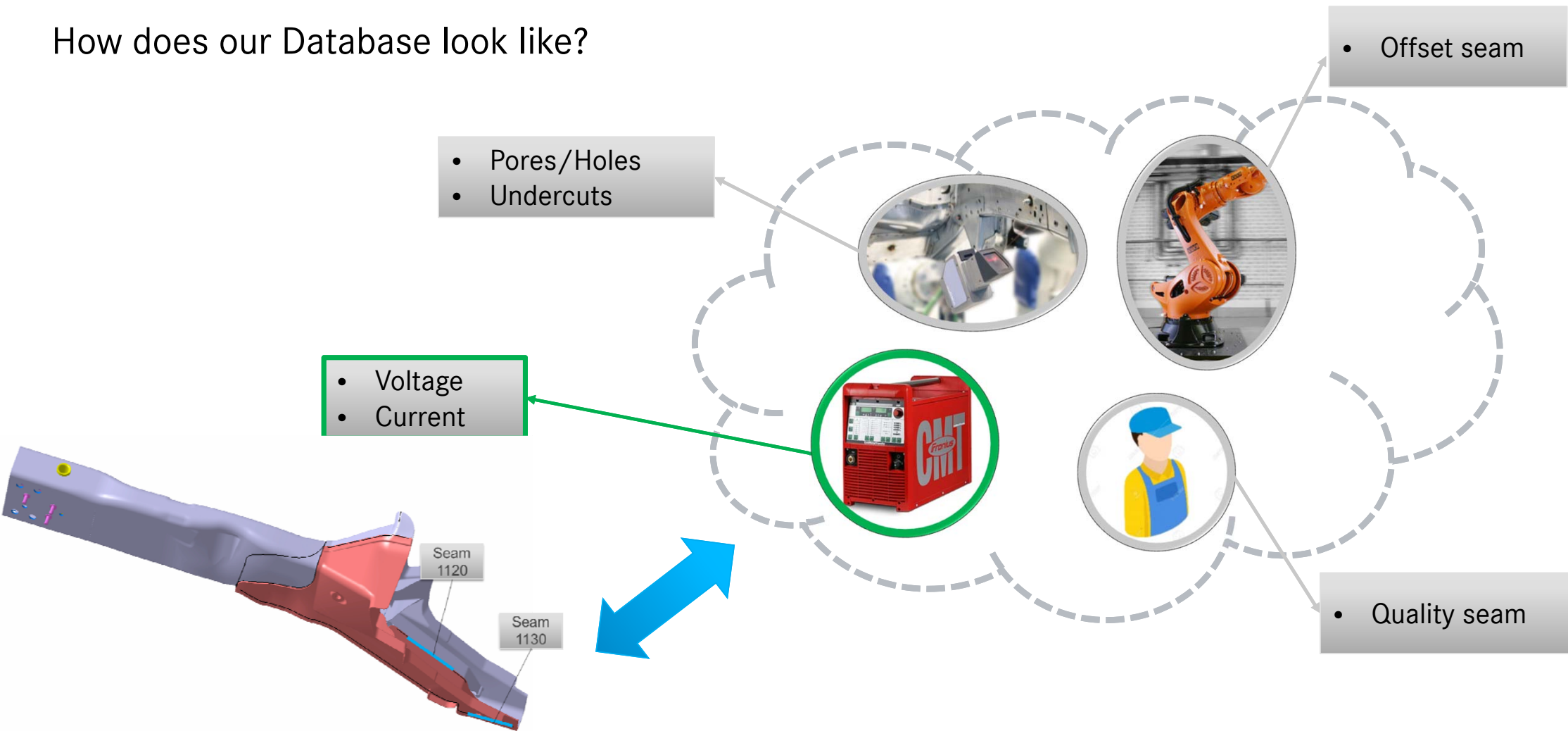


The „modern“ gas metal arc welding process is a mixture of different process types

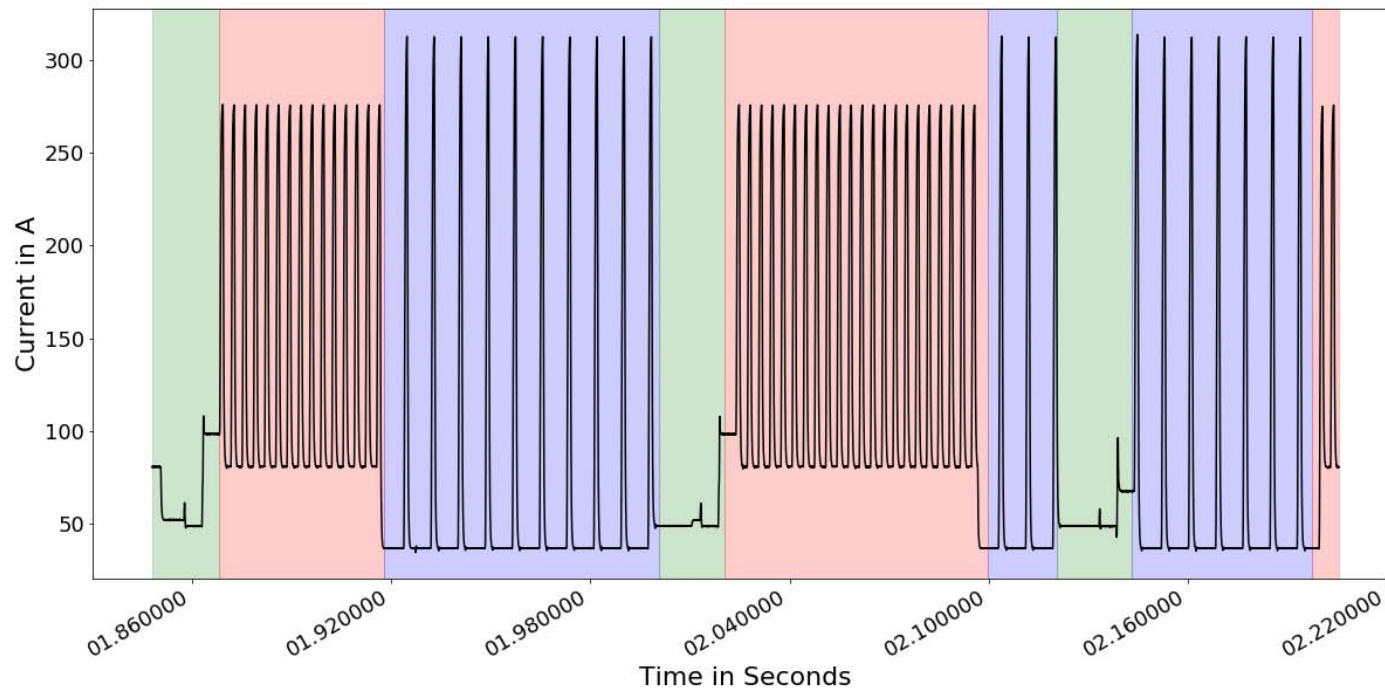


- Controlable heat input
- Stable arc length
- Better gap bridgeability
- Evenly rippled surface

How does our Database look like?



Procedure: From process data to weld seam quality



Automatic
segmentation of
similar process
cycles

Segmentation

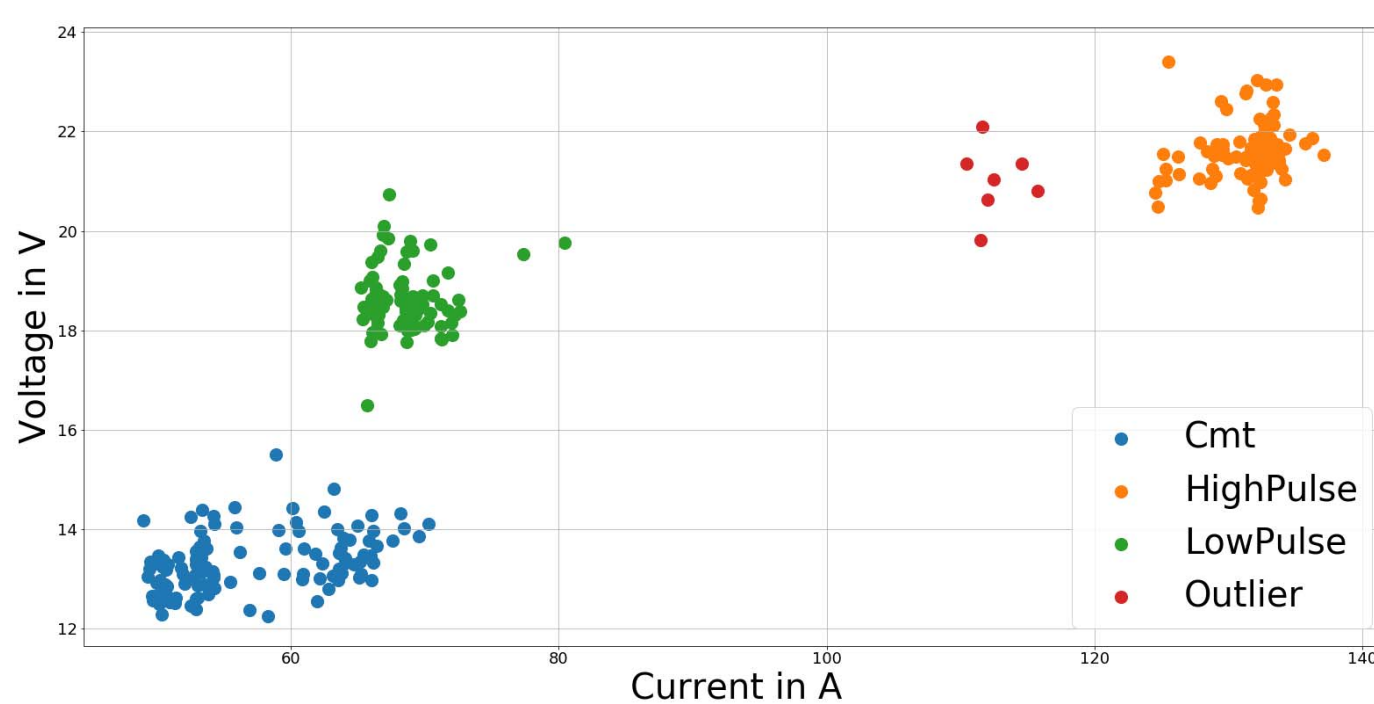
Classification

Extraction

Correlation

Result

Procedure: From process data to weld seam quality



Classification of the segments with unsupervised learning

Segmentation

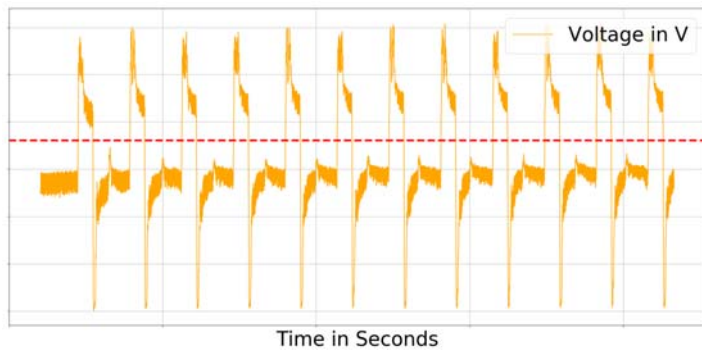
Classification

Extraction

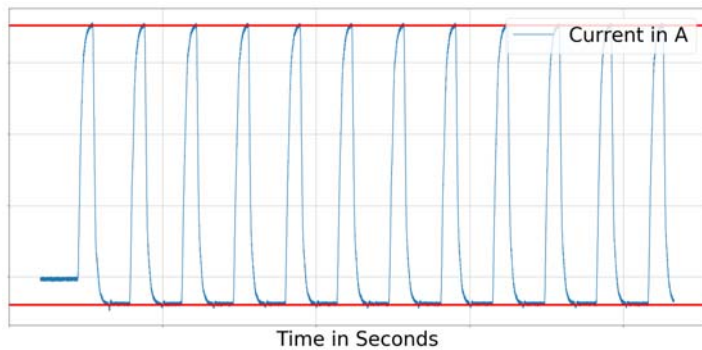
Correlation

Result

Route: From process data to weld seam quality



- Mean
- Variance
- Skewness
- Kurtosis



- Base time
- Pulse time
- Base current
- Pulse current

Extraction of statistical and process specific features

Segmentation

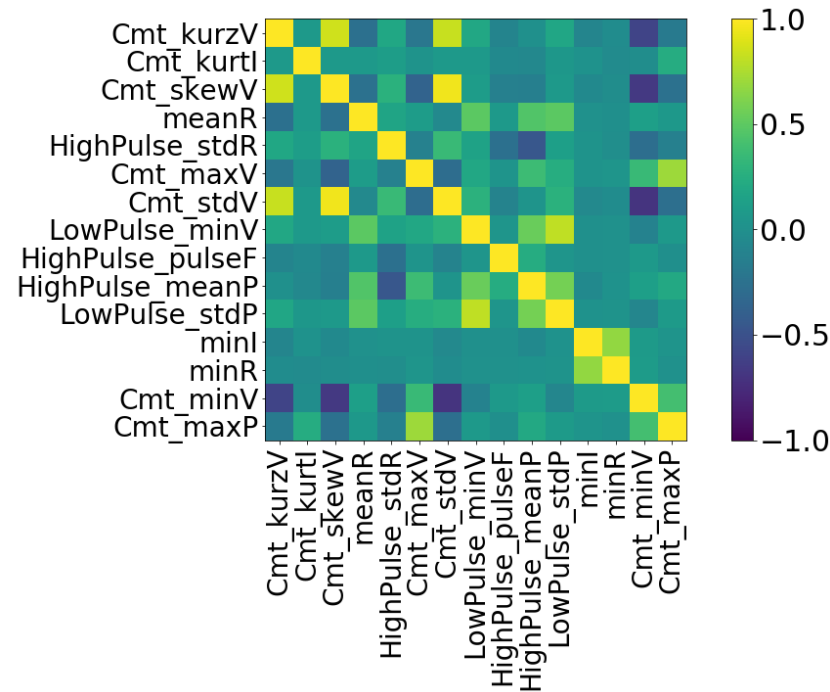
Classification

Extraction

Correlation

Result

Procedure: From process data to weld seam quality



Analysis of correlation between process and quality features



Basics: The correlation coefficient measures the relation between features and target



Consumption [ltr/ 100km]

10

15

Time [min]

60

45

Color

yellow

blue

Add-ons

climatronic

bending lights

Target

Who was speeding?



Basics: The correlation coefficient measures the relation between features and target



Consumption [ltr/ 100km]
Time [min]
Color
Add-ons

High
High
Low
Low

Correlation-coefficient



Who was speeding?

Basics: The correlation coefficient measures the relation between features and target



Consumption [ltr/ 100km]

10

15

Time [min]

60

45

Color

Pink

Blue

Add-ons

parking
assistance

bending lights

Target

Which car was driven by a „hipster“?



Basics: The correlation coefficient measures the relation between features and target



Consumption [ltr/ 100km]

Time [min]

Color

Add-ons

Correlation-
coefficient

Target

Which car was driven by a „hipster“?



Basics: The correlation coefficient measures the relation between features and target



Consumption [ltr/ 100km]

Time [min]

Color

Add-ons

Low

Low

High

High

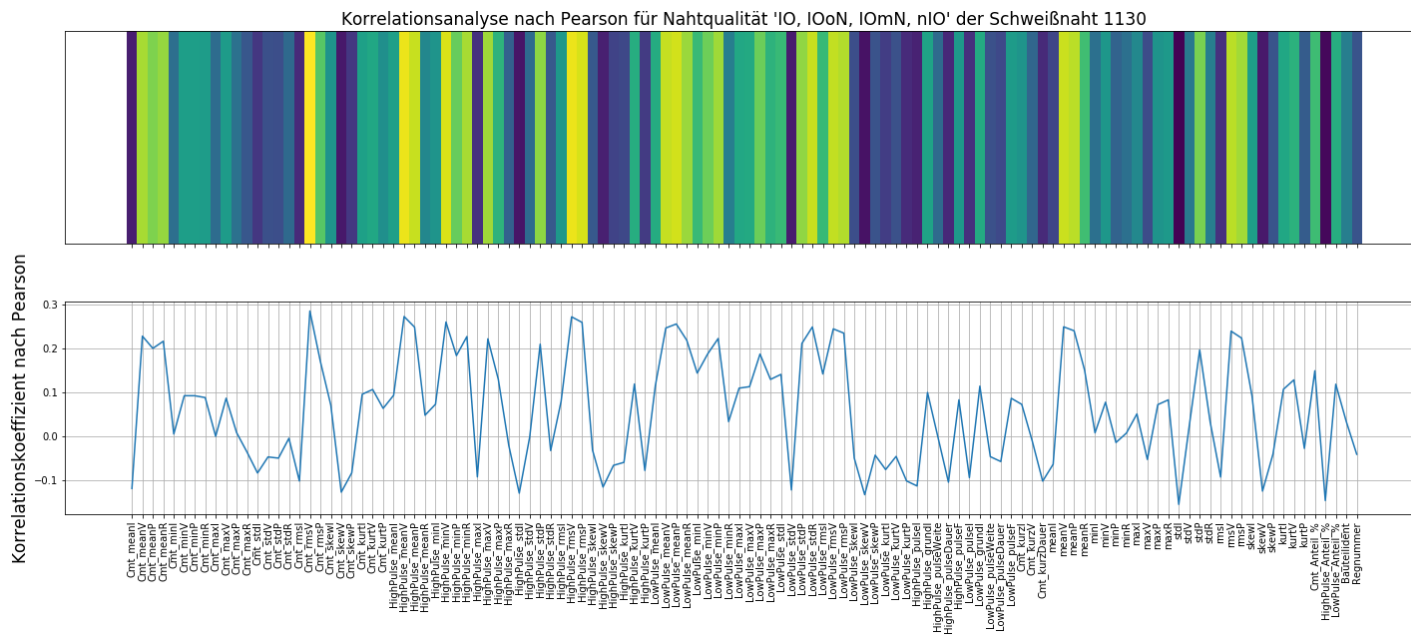
Correlation-
coefficient



Target

Which car was driven by a „hipster“?

Procedure: From process data to weld seam quality



Selection of high coefficient

Segmentation

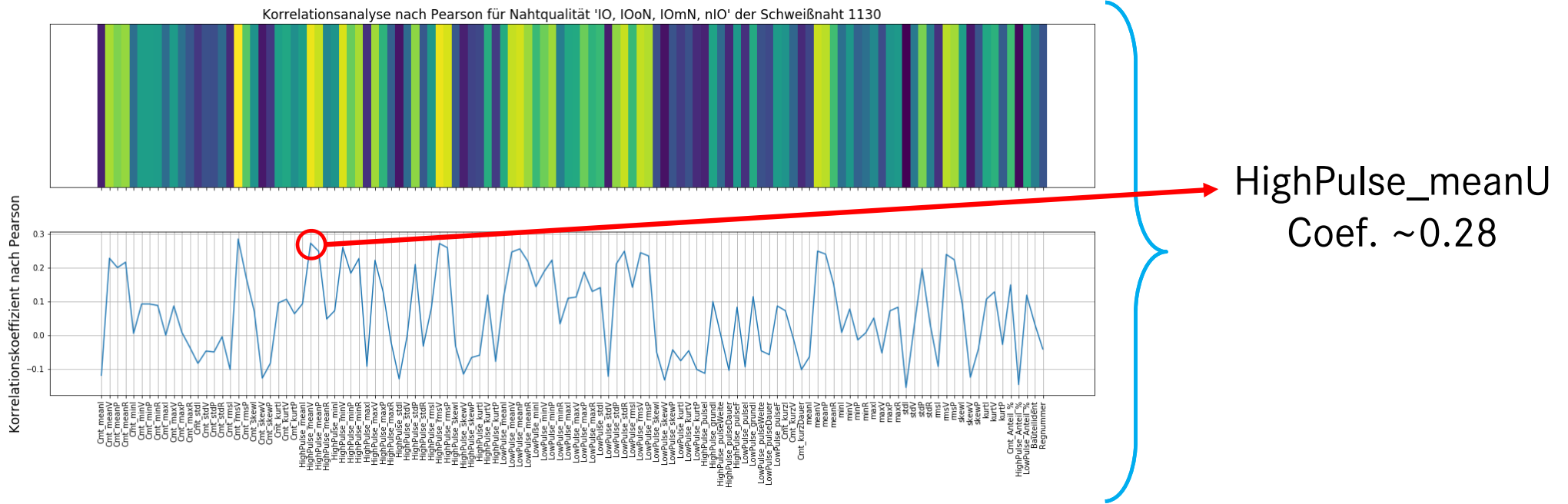
Classification

Extraction

Correlation

Result

Procedure: From process data to weld seam quality



Segmentation

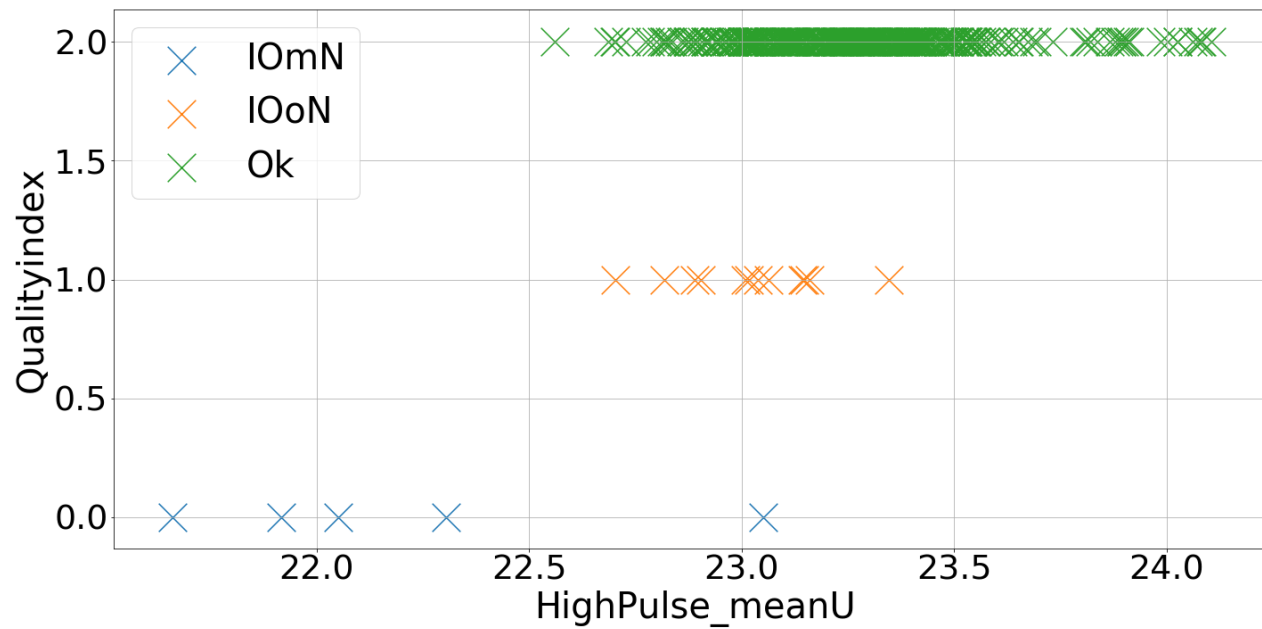
Classification

Extraction

Correlation

Result

Procedure: From process data to weld seam quality



HighPulse_meanU
Coef. ~ 0.28

Segmentation

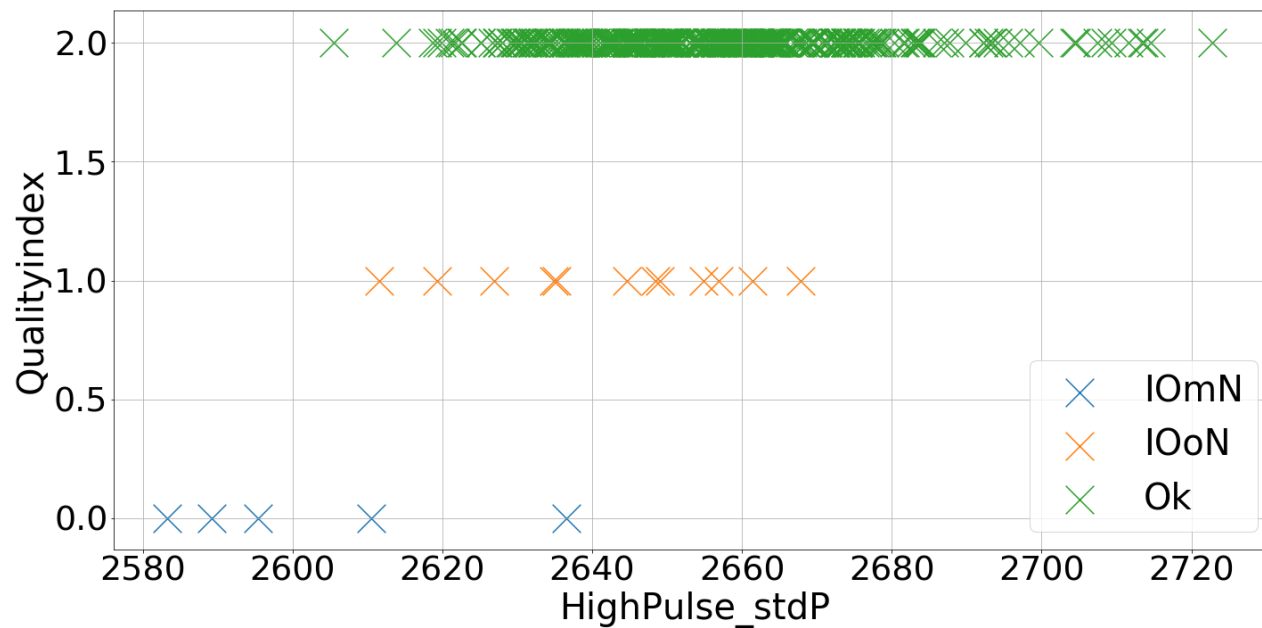
Classification

Extraction

Correlation

Result

Procedure: From process data to weld seam quality



HighPulse_stdP
Coef. ~ 0.23

Segmentation

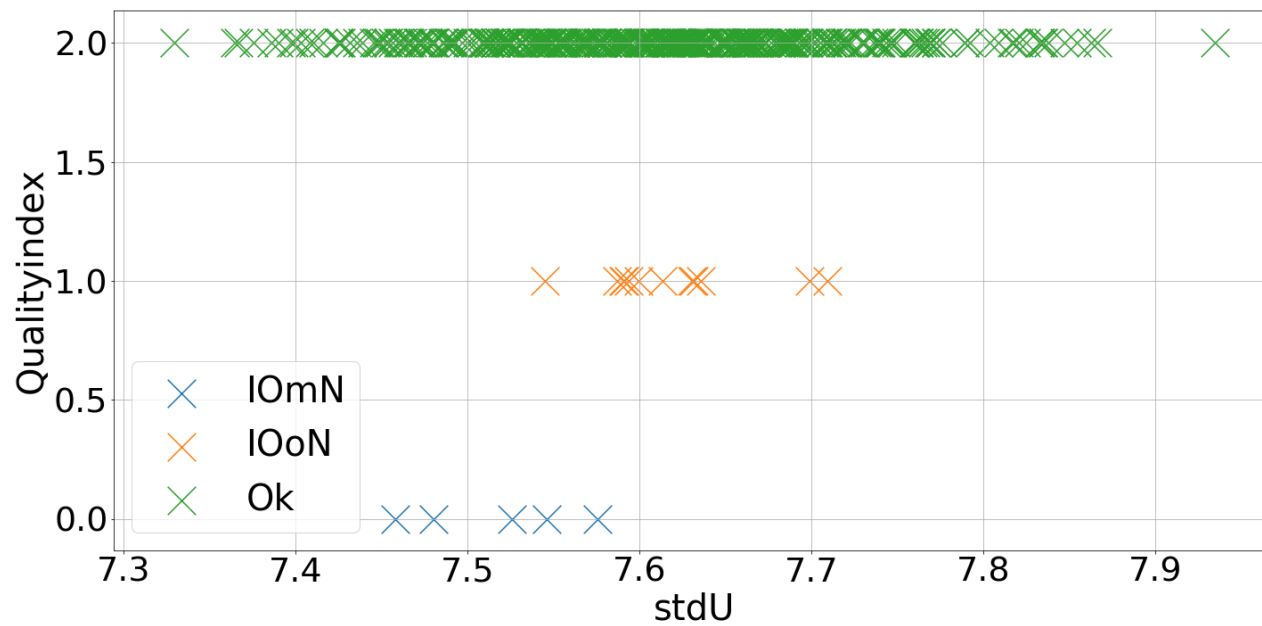
Classification

Extraction

Correlation

Result

Procedure: From process data to weld seam quality



stdU
Coef. ~ 0.06

Segmentation

Classification

Extraction

Correlation

Result

Summary

First steps are done- next ones have to follow

- Feasibility studies
- Features in time domains
- Criteria: IO, IOoN, IOmN



Summary

First steps are done- next ones have to follow

- Feasability studies
- Features in time domains
- Criteria: IO, IOoN, IOmN



- Statistical significance
- Features in frequency domains
- Determination of error types

Outlook

First steps are done- next ones have to follow

- Feasability studies
- Features in time domains
- Criteria: IO, IOoN, IOmN



- Statistical significance
- Features in frequency domains
- Determination of error causes



Mercedes-Benz

Bigger data basis, more classified welding errors

Thank you very much for your attention!

