# PRESED -

# Predictive Sensor Data Mining for Product Quality Improvement

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# **Project Overview**

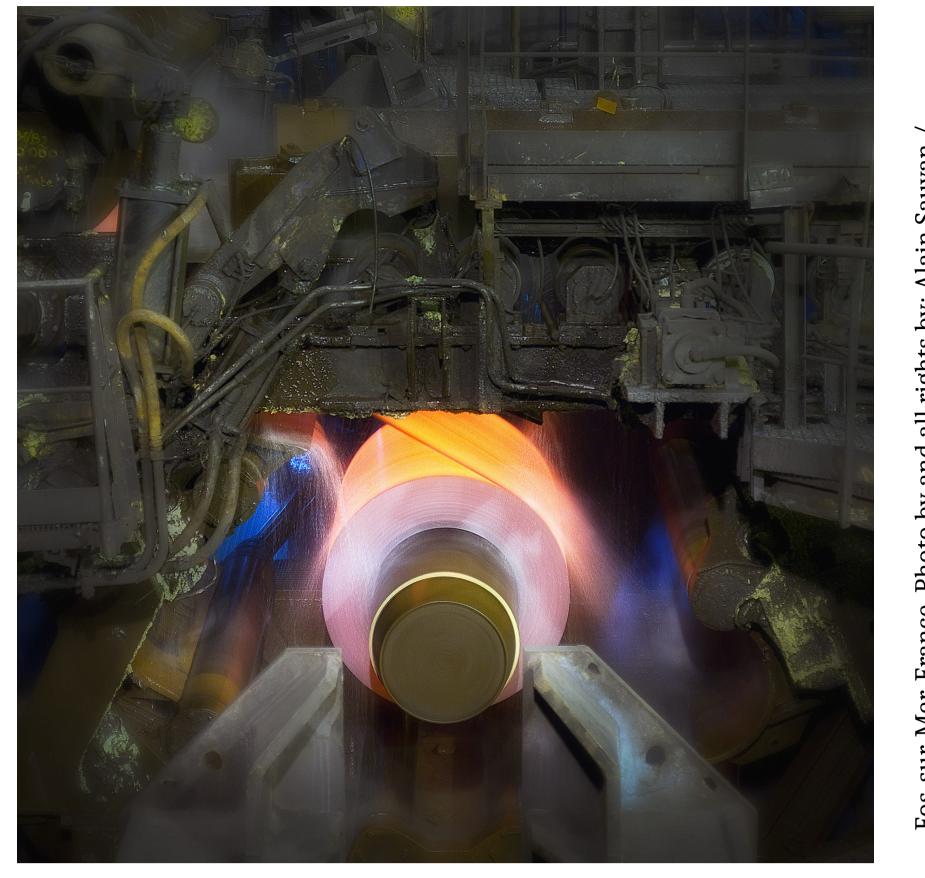
Manufacturing high quality steel products is a very complex process with a lot of intermediate steps.

Modern steel plants also produce huge amounts of complex data from sensors and production parameters.

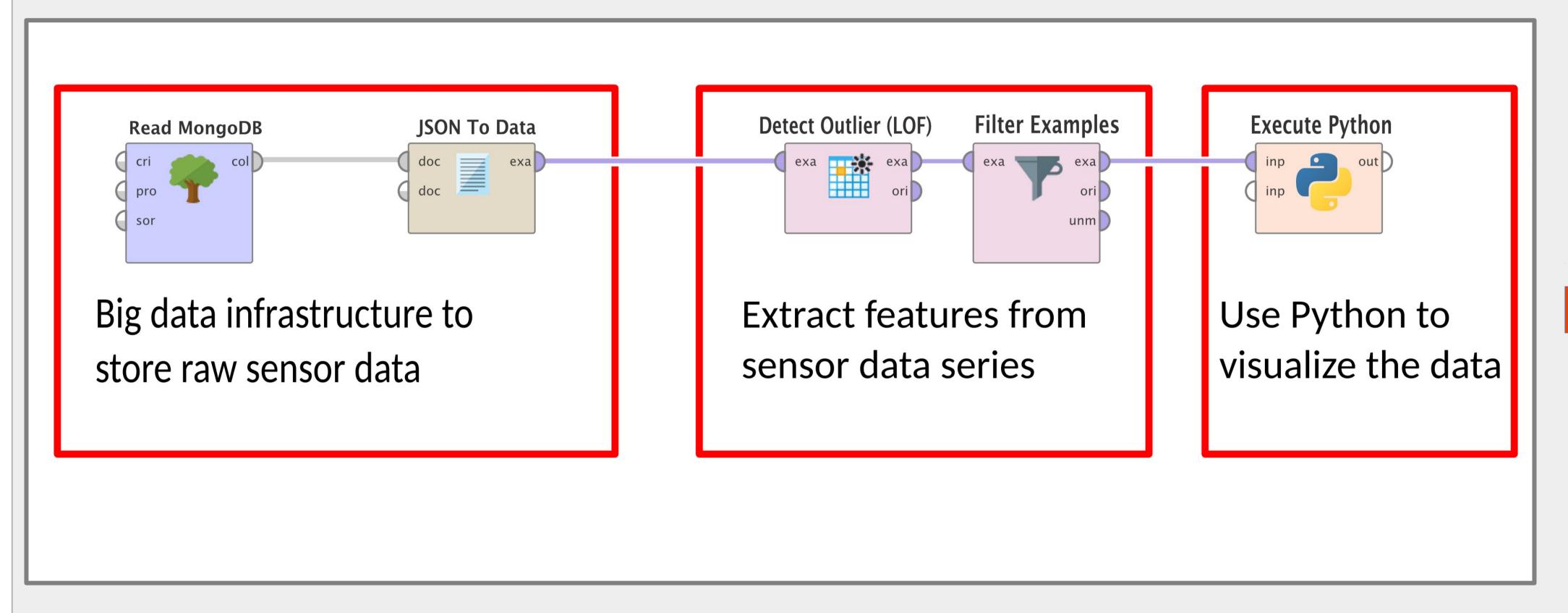
The goal of our project is to develop new methodologies to help steel plants to improve their products and reduce manufacturing costs.

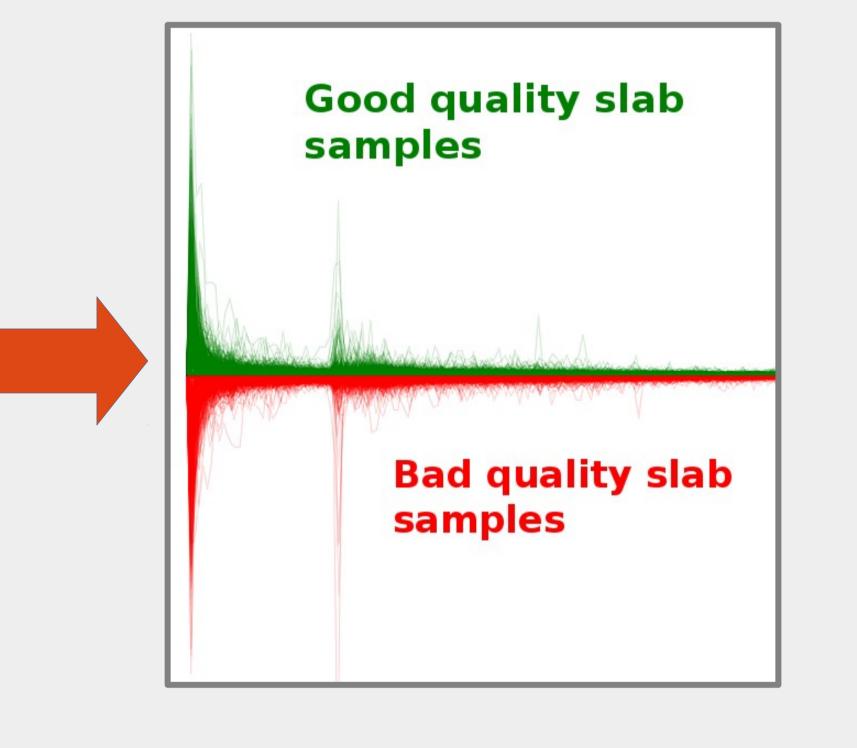
#### **Use Cases**

- Identifying causes for low product quality
- Predict the quality of a product during production as early as possible
- Integrate expert knowledge for different production sites



os-sur-Mer, France. Photo by and a





### RapidMiner process showing the access and transformation of the raw data

## Overlay of Fourier transformed signals

## Challenges

- Design a data structure to track the material over the complete production process
- Tracking is difficult as the product changes shape and quantity during processing (rolling, cutting)
- Huge amount of raw data (several hundred parameters, with a frequency of 1-10Hz over a 2-3 year period)

# Continuous Caster Dynamic variables / Time series Argon Flow Labels Hot Rolling Mill Static variables / Time series Wavelet transform of casting speed Wavelet transform of casting speed Wavelet transform of casting speed Labels Raw Data UC2 Hot roll mill Cold roll mill Pickling line Raw Data UC3 Electric arc furnace Continuous casting

### Generic data model for the PRESED use cases

# Technologies used

- MongoDB as NoSQL database to handle the production data
- Data visualization tool box written in *Python* and integrated into *RapidMiner* [1]
- Representation learning to process time series information [2]
- Time series shapelets for pattern discovery [3]
- Outlier detection methods for time series [4]
- Data and model management with RapidMiner
- Dashboards for interactive exchange between plants and data mining experts
- Ontology management with *KASEM* [5]

### References

- [1] Ingo Mierswa, Michael Wurst, Ralf Klinkenberg, Martin Scholz, and Timm Euler. 2006. YALE: rapid prototyping for complex data mining tasks. In Proceedings of the 12th ACM SIGKDD international conference on Knowledge discovery and data mining (KDD 2006)
- [2] A. Ziat, G. Contardo, N. Baskiotis, and L. Denoyer, "Car-Traffic Forecasting: A Representation Learning Approach.," MUD@ICML, pp. 85–87, 2015
- [3] Xavier Renard, Maria Rifqi, Walid Erray, and Marcin Detyniecki. Random-shapelet: an algorithm for fast shapelet discovery. IEEE International Conference on Data Science and Advanced Analytics, pages 1–10, 2015
- [3] Cateni Silvia, Valentina Colla, and Gianluca Nastasi. "A multivariate fuzzy system applied for outliers detection." Journal of Intelligent & Fuzzy Systems 24.4, 2013
- [5] Monnin, M, Leger, J-B., Morel, D. (2011). KASEM®: e-Maintenance SOA Platform, in Proceedings of 24th International Congress on Condition Monitoring and Diagnostics Engineering Management, 29th May 1st June, Stavanger, Norway

# **Project partners**









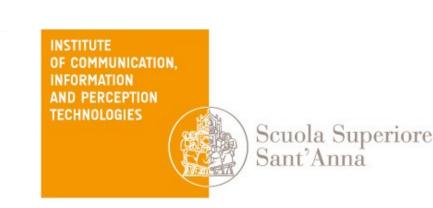
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