

Failure prediction and process monitoring using Machine Learning at MONDI Gronau

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Mondi Gronau GmbH







Jöbkesweg 11 48599 Gronau, Deutschland

Local: Global:

~ 850 employees ~ 25.000 employees

Mildenberger & Willing -> Nordenia -> Mondi

Production Volume:	170	Mio. kg
Waste Volume:	15	Mio. kg
Number of rolls:	1.7	Mio. Stk.
Yield:	421	Mio. €
Energy Consumption:	71	Mio. kWh
Production Time:	24/7	hh/dd
Square meter:	104	k. m²

Facts & Figures: References





Johnson Controls







P&G



Facts & Figures: Machines



Extrusion Lines



- Monoextrusion
- Coextrusion
- Film thickness:
- Film width:

- n Layer
- 10 300µm
- 850 3.000mm



Facts & Figures: Machines



Processing Types

Coating / Siliconizing



Slitting



Rotogravure Printing



Lamination



Facts & Figures: Products



Benefits Mondi Gronau GmbH

Good Product



Waste Product



Benefits

- Advanced Quality Monitoring
- Reduction of Waste Material
- Customer Satisfaction
- Yield Optimization



Monitoring Systems

Monitoring: In-Line



Integrated Monitoring System



Optical Control System

- Laminated Film Surface Detection
- Traffic Light System
- Quality Index
- Addional Systems: Colour, Thickness



Monitoring: PLC Data





Parameters/ Features (100 – 500)

PLC

- PLC (Programmable Logic Controller), Data Collector
- 4-5 PLC per machine for real-time acquisition



Data Processing







- Acquisition
- Pre-Processing (ETL)
- Limits/ Targets from Customer Specifications
- Visualization On-Line/ Off-Line



Next Step in using Prediction Methods



Separated Dosing/ Pressure Datasets Machine M150 Subsection (7.1.1) Data (7.1.2) Validation (7.1.2) Training Off-line Off-line **Results Examination** (7.1.4) Online Novelty (7.1.4) OCC Prototype (7.1.3) Testing Transfer into Detection Abnormal 2D by MDS Boundary oundary Normal **On-line On-line** Off-line to new data A: MDS Scaled View 6-Dimensional **C: Threshold Setting** for Zone Visualization (Sammon Mapping) 2-Dimensional B: Delaunay/ Convex Hull

Acquisition Loop

- Acquisition
- Pre-Processing (ETL)
- Machine Learning Methods/ Models
- Visualization On-Line/ Off-Line





Next Step in understandable Visualization



Up to 200 parameters in one point [temperature, pressure, speed,...] Acquired per minute Stored on an Oracle database Processed for visualization in lower dimensions

Visualization

Reduction of information to understandable level (1, 2, 3 dimensions)
Visualization in real-time



Version 3.0 (Mathworks)

Next Step in Software Development

Version 1.0 (internal)



Acquisition Loop

- Acquisition of more Datasets
- Pre-Processing (ETL)
- Extended Machine Learning Methods/ Models
- Version 3.0 of Visualization On-Line/ Off-Line



Human-Machine-Interface

Human-Machine-Interface: Industrie 4.0



Processing Loop



Processiveringahl v1.5 @ Mendi Groney GmbH 201

Application requirements

Retrieve, analyze and visualize machine data

- Up to 40 machines with up to 500 sensors
- Updated once per minute near real time
- Alarm events and error logging
- Intuitive user interface
- High robustness
- Expandability
- Failure forecasts for increased quality / downtime reduction





Application requirements

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Up to 40 machines with up to 500 sensors

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Expandability



Failure forecasts for increased quality / downtime reduction



Prozesskennzahl v3.0 / Key features





User Interface



 Current machine status

 Visualization for up to 72 hours



Summarized Info

Visualize sensor data

 Limits to trigger alarms and warnings

Forecast analysis

Main status

Plug-In feature





- Add new machines without code changes
- Customized calculation and visualization per machine
- Code for plug-in and main application separated





Next:

Algorithms Software

Process Monitoring Algorithms and Software





Which sensor measurements indicate machine failure?



Process Monitoring Algorithms and Software

Basic Workflow





Process Monitoring Algorithms and Software Pre-Processing



Sensor data and quality states are aggregated (per time stamp)



Process Monitoring Algorithms and Software -Train a prediction model

Basic Workflow





Process Monitoring Algorithms and Software– Train a prediction model

Possible Classification Methods

Statistics and Machine Learning





Process Monitoring Algorithms and Software -Train a prediction model

Basic Workflow





2

Process Monitoring Algorithms and Software– Train a prediction model

Fit model based on historic data

PredictionModel = fitctree PARAMETER, STATE) 2 TIMESTAMP STATE PARAMETER **Training Data** '2015-07-14 00:49:12.0' 160 160 160 1000 7 1000 33 160 9 32 e.g. 60% of '2015-07-14 00:50:12.0' 160 160 160 1000 8 1000 10 33 32 160 '2015-07-14 00:51:13.0' 160 160 160 1000 8 1000 10 33 32 160 historic data 32 '2015-07-14 00:52:12.0' 160 160 160 160 1000 8 1000 10 33 (3 months) 32 11 33 '2015-07-14 00:53:12.0' 160 160 160 1000 8 1000 160 '2015-07-14 00:54:12.0' 160 160 1000 8 1000 12 33 32 160 160 '2015-07-14 00:55:12.0' 160 160 160 1000 8 1000 10 33 32 160



Process Monitoring Algorithms and Software -Train a prediction model

Basic Workflow





Process Monitoring Algorithms and Software – Train a prediction model



Misclassification rate 1 of 7: 14.28 %



Process Monitoring Algorithms and Software –

Basic Workflow





Process Monitoring Algorithms and Software -Application

Predict current machine states during operation





Process Monitoring Algorithms and Software -Application





Thank you!

Questions?