

Norkshop Al in Photonics Swissmem und Swissphotonics

Machine learning for optical quality inspection

Thomas Grünberger plasmo Industrietechnik GmbH Vienna 2019-09-03



content

- company presentation
 plasmo
- o data is not information
 - o human intelligence
 - o artificial intelligence
- examples machine learning









company plasmo



global. focussed. independent.









800

more than 800 plasmo systems in operation













100

more than 100 global customers using plasmo





WE MAGNETISE THE WORLD



plasmo quality suite visualization









basic technologies





plasmoeye

06



contact

austria

plasmo Industrietechnik GmbH Dresdner Str. 81-85 1200 Vienna phone +43 1 236 26 07-0 www.plasmo.eu

germany

plasmo Industrietechnik GmbH NL Deutschland Leitzstr. 45 70469 Stuttgart phone +49 711 49066-307 www.plasmo.eu



usa

plasmo USA LLC 44160 Plymouth Oaks Blvd, Plymouth, MI 48170 phone +1 734 414 7912 www.plasmo-us.com

china

plasmo China 42F Wheelock Square 1717 Nanjing West Road Shanghai 200040/P.R. China phone +86 21 8028 6166 www.plasmo.cn



data is not information



technology

strategy

VISION

INNOVATION

creativity



how to extract information from data

- o data plausibility
- data analytics 0
 - o human intelligence
 - o artificial intelligence
 - o machine learning
 - o deep learning







data analytics

- o data architect
- o data engineer
- o data scientist
- o business analyst
- \circ devops





SQL/no SQL

o Docker

• Kybernets

o DFS

(i . . .

0

cloud/edge/fog



human intelligence and graphical data analytics aggregation seam and part level



All Data records: 32152; Current selection: 0 data records (0.00 %); Context: 0 data records (0.00 %); Focus: 0 data records (0.00 %)



·•	DateTime 🛛 🖌	Index 🛛 📕	channel_ok_asString	StationName 🛛 📉	PartType 📕	SeamName 🛛 📉
0-	2015-03-19 00:00:00	1	ОК	Machine1	Part1	Part1N1
	2015-03-19 00:00:00	2	ОК	Machine1	Part1	Part1N2
0	2015-03-19 00:00:00	3	ОК	Machine1	Part1	Part1N3
0-	2015-03-19 00:00:00	4	ОК	Machine1	Part1	Part1N4
	2015-03-19 00:00:00	5	ОК	Machine1	Part1	Part1N5
:57	2015-03-19 00:00:00	6	ОК	Machine1	Part1	Part1N8
	2015-03-19 00:00:00	7	ОК	Machine1	Part1	Part1N7
	2015-03-19 00:00:00	8	ОК	Machine1	Part1	Part1N6
0	2015-03-19 00:00:00	9	ОК	Machine1	Part1	Part1N9
0	2015-03-19 00:01:00	10	ОК	Machine1	Part1	Part1N1
	2015-03-19 00:01:00	11	ОК	Machine1	Part1	Part1N2
0	2015-03-19 00:01:00	12	ОК	Machine1	Part1	Part1N3
	2015-03-19 00:01:00	13	ОК	Machine1	Part1	Part1N4
	1	10000 32152	OK NOK	Machine1 Machine2	Part2	52
.57						

3



human intelligence and graphical data analytics aggregation machine level







human intelligence and graphical data analytics aggregation site level



All Data records: 32152; Current selection: 0 data records (0.00 %); Context: 0 data records (0.00 %); Focus: 0 data records (0.00 %)



				plasmo
				14000
				12000
				10000 (jo 3X 10000
				Value (Bo
	•			Mean
				6000
				4000
	-			2000
1	StationName	Machine2	🐼 Au	to 📪 📑



human intelligence and graphical data analytics root cause analysis









human intelligence and graphical data analytics process optimization

- statistical analysis of
 - defect positions
 - o defect types
 - o date/time
 - material/vendor
 - maintenance planning
 - Ο . . .
 - finding correlations, Ο trends









All Measurements: 698; Current selection: 345 measurements (49.43 %); Context: 0 measurements (0.00 %); Focus: 0 measurements (0.00 %)



artificial intelligence





artificial intelligence definitions

- o Al Artificial Intelligence
 - tries to model human intelligence
 cybernetics, ...
- o statistics
 - tries to define what happened
 DM and ML came from statistics



- DM Data Mining
 - tries to explain why something happens (e.g. root cause)
- o ML Machine Learning
 - tries to explain what will happen in future and how to optimize or avoid certain situations
 - ML is a first step for model human intelligence and can be seen as part if AI



machine learning techniques

- o supervised learning
 - o target is known
 - develop model based on input and output data
- o unsupervised learning
 - group and interpret data based only on input data







examples



diode based process monitoring welding of thin sheets

- 2 sensors are used (different wavelengths), 710 test runs (index), one measurement consists of 2050 measurement values.
- for one complete seam (4 dimensions)







information is expected in the characteristics mean value and standard deviation





diode based process monitoring unsupervised learning

are there clusters in the data? 0

k-means





manifold learning





diode based process monitoring supervised learning

OKNOK and defect types were analysed via DT and NDT techniques

OKNOK distribution





classification logistic regression





diode based process monitoring supervised learning

Ο yields in optimal results (comparison based on confusion matrix)







Actual



comparison of different modelling techniques for OKNOK, ANN and random forest

Actual

random forest

artificial neural network



Predicted





diode based process monitoring supervised learning

○ 3D visualisation of the task (OK runs type 1 and 3 defect types (-3, -2, -1)









diode based process monitoring summary

unsupervised learning shows 4 different 0 clusters correlating with the 4 defect types

supervised learning using machine learning 0 techniques enables correct classification



at least 3 different characteristics and at least 2 different diode signals needed





camera based weld seam monitoring

- $\circ~$ camera based weld seam monitoring enables the detection of visible defects
- used in addition to process monitoring systems detecting defects defects like lack of fusion, porosity, no full penetration, ..., which can't be seen at the top of the seam
- o gray value or color images useable
- example welding of C shaped seams







camera based weld seam monitoring supervised training, deep learning

- deep learning are available the last few years due to computational power 0 (e.g. GPU processing)
- artificial neural networks are used typically using many neurons
- different approaches available including pretrained nets for specific tasks Ο
- unsupervised deep learning also usable
- supervised learning using training, test and validation data set Ο

 check for inter- and extrapolation capabilities (e.g. overtraining) plasmo



camera based weld seam monitoring supervised training, deep learning

trainings data set 0 classified correct









275.jpeg|0:0, trained











279.jpeg|0:0, trained











316.jpeg|0:0, trained





319.jpeg|0:0, trained





320.jpeg|0:0, trained





camera based weld seam monitoring supervised training, deep learning

test data set 0 1 false positive and 1 false negative



























NIO NIO 100%



camera based weld seam monitoring supervised training, deep learning

trainings data set
 classified correct





Image 026.png|0:0, trained







camera based weld seam monitoring supervised training, deep learning

test data set classified correct











Image 017.png(0:0



| p



camera based weld seam monitoring summary

- deep learning gives better results compared to manual inspection 0
- unsupervised approach also applicable (presentation of OK seams and detection Ο of anomalies)
- models can be trained once or retrained online
- result depends on input data and correct classifications (supervised) Ο







machine learning at plasmo summary

- $\circ\,$ dashboard based analysis of production and sensor data (SPC)
 - o correlations
 - o trend analysis
- modelling OKNOK for different sensors for quality inspection of different joining techniques (laser, TIG, plasma, FSW, ...)
- unsupervised clustering of sensor data for all applications
- genetic algorithms for supervised or unsupervised automatic parameterisation of quality inspection systems





machine learning at plasmo summary

- deep learning for image based analysis Ο
 - analysis of powder bed images (AM) and classification of different Ο defect types
 - Ο





unsupervised analysis of image staples (layer by layer) in additive manufacturing

example 3D visualisation of 50 layers building a bridge (1mm height)

-> up to 10.000 layers (images) has to be analysed for a real job





machine learning at plasmo summary

- some tips from our experience 0
 - don't learn existing knowledge
 - keep it as simple as possible
 - 80percent rule: 80% is data preparation, 20% is data analysis Ο





Tony D'Amato (Al Pacino) in "Any given Sunday"

"Inch by inch. Play by play. Until we get there."

