

DATA SCIENCE COMPLEXITY AND SOLUTIONS IN REAL INDUSTRIAL PROJECTS

ROSEN

empowered by technology

WHO AM I

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Slides: <https://github.com/rosen-group/conferences>

INTRODUCING THE ROSEN GROUP



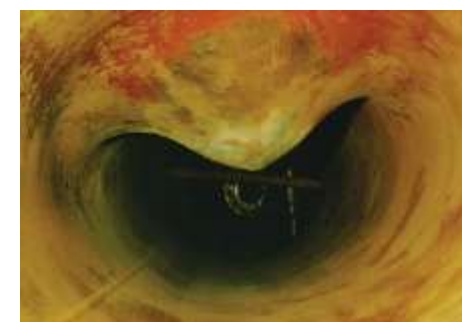
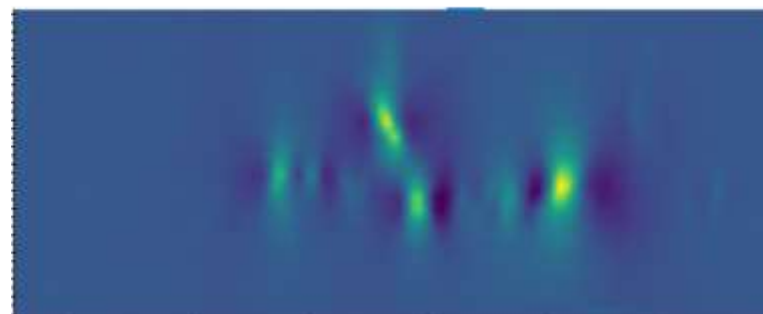
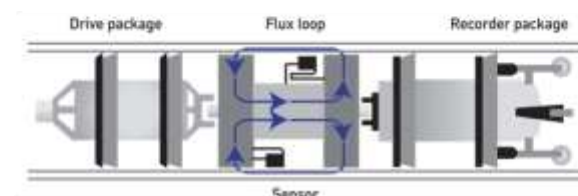
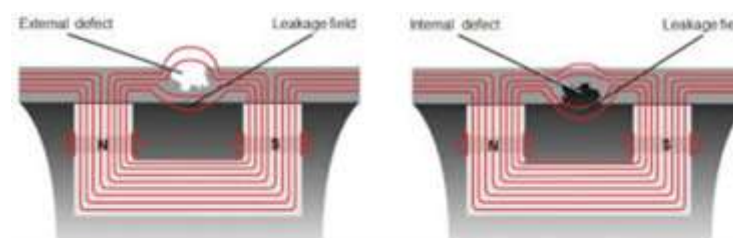
ROSEN develops and manufactures equipment, software, and methods for the inspection, diagnosis, and protection of industrial structures in a wide range of industries.

Because damage can cause serious impacts!

CHALLENGES

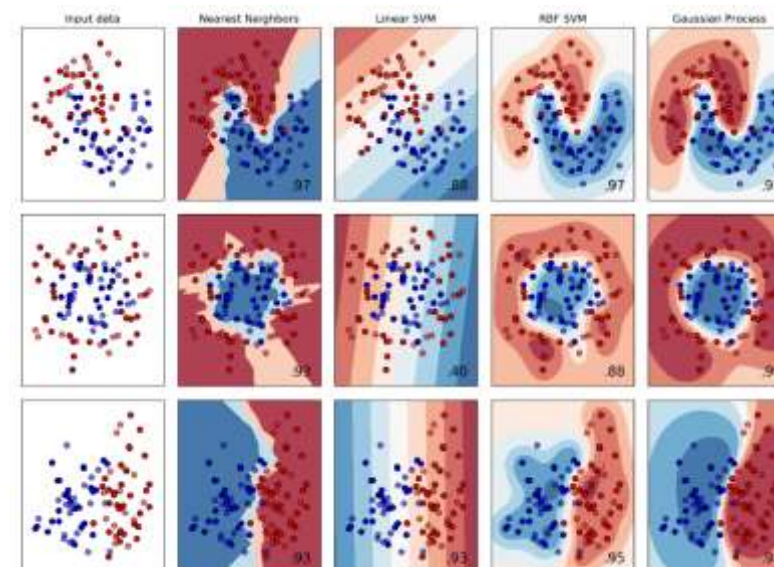
- **Classification** of installations and anomalies and accurate **estimation** of severity of defects.
- Our tools record **a huge amount data**, up to multiple terabytes per run.
- Severe defects threaten the integrity of the pipelines, therefore there is a **high risk** for environment and clients.

We address these challenges with **machine learning** and **Python**!



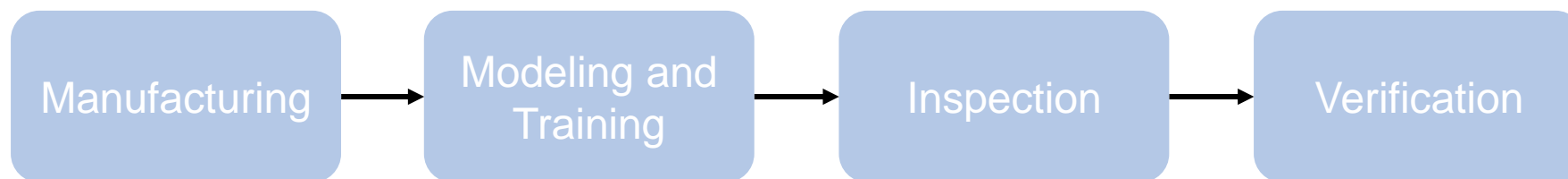
DATA SCIENCE MISCONCEPTION

- Misconception: Most of the time is spend tweaking machine learning models
- Real data is much more complex than toy datasets
- What is missing?
 - Data gathering
 - Data cleaning
 - Missing and imbalanced data
 - Scaling



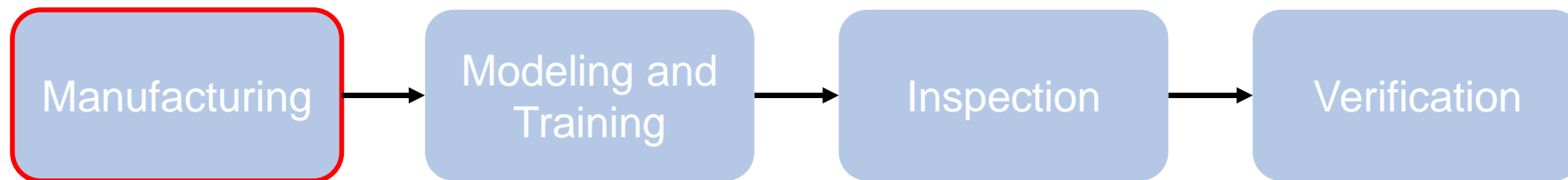
CLASSIC APPROACH

- Manufacturing and calibration of tools
- Models are based on pull-tests and laboratory measurements
- Verification results aren't fed back into the models



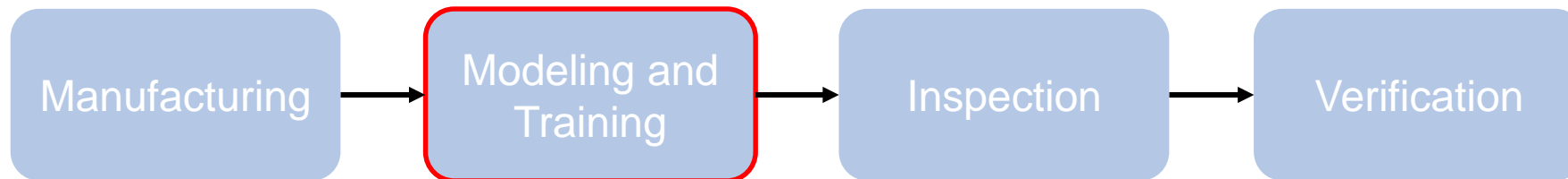
MANUFACTURING

- Tool is manufactured
- Laboratory measurements
- Pull-test or Pump-test
- Calibration
- Artificial defects



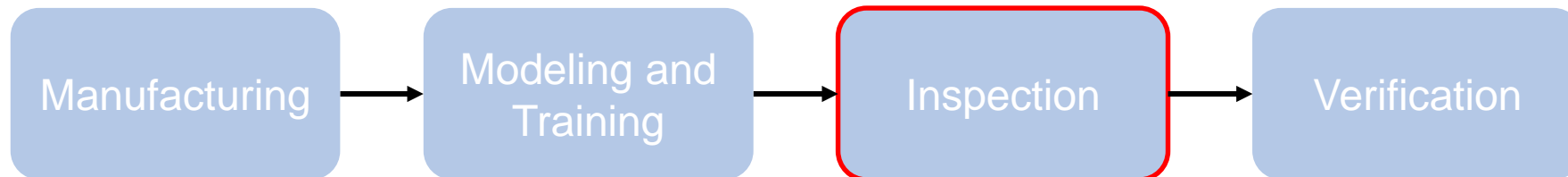
MODELING AND TRAINING

- Estimating the depth and shape of defects with regression models
- Classification of installations and anomalies
- Data from
 - Pull-tests and pump-tests
 - Laboratory measurements



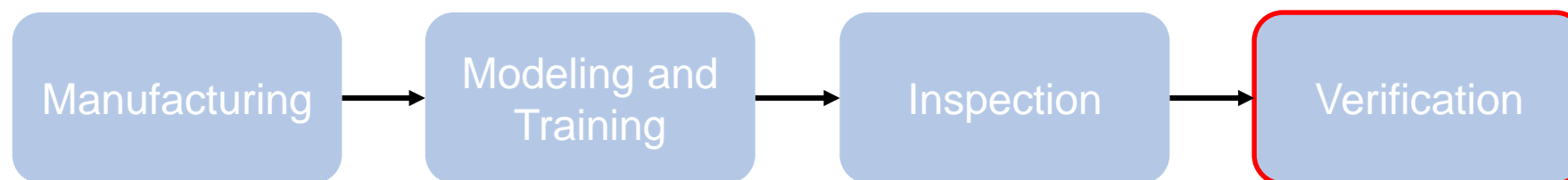
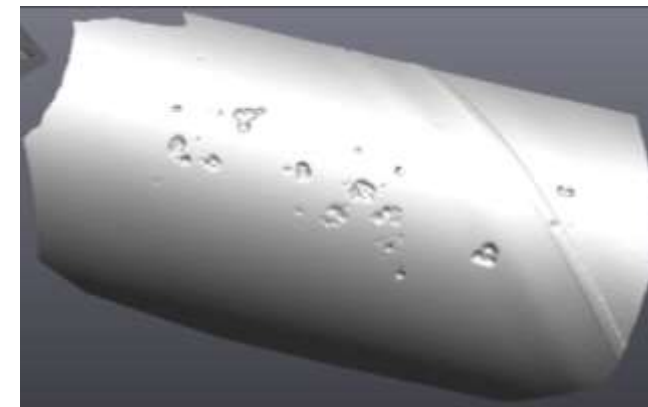
INSPECTION

- Preparing, transporting and launching the tool
- Tool takes measurements of the pipeline
- Processing and analyzing the data
- Reporting



VERIFICATION

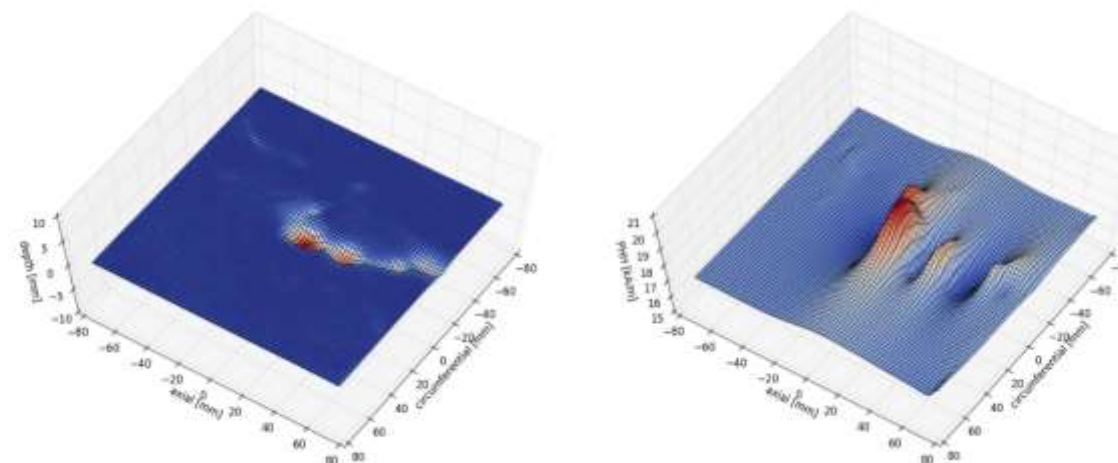
- Dig-ups
- Only severe defects
- Laser-scans
- X-ray computed tomography



VERIFICATIONS

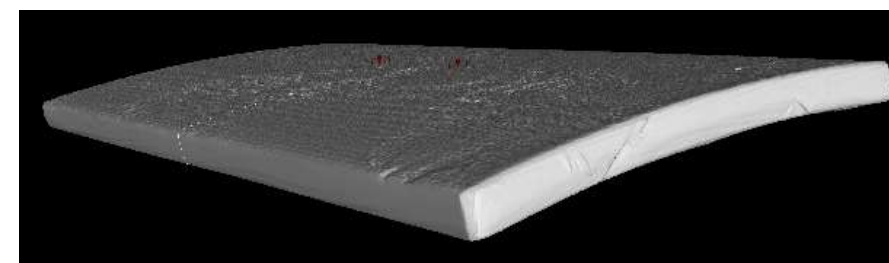
Laser-scans

- High resolution image of the outer pipeline wall
- Good image of corrosion
 - Depth and shape
- Better than artificial defects (e.g. ellipses)
- Non destructive



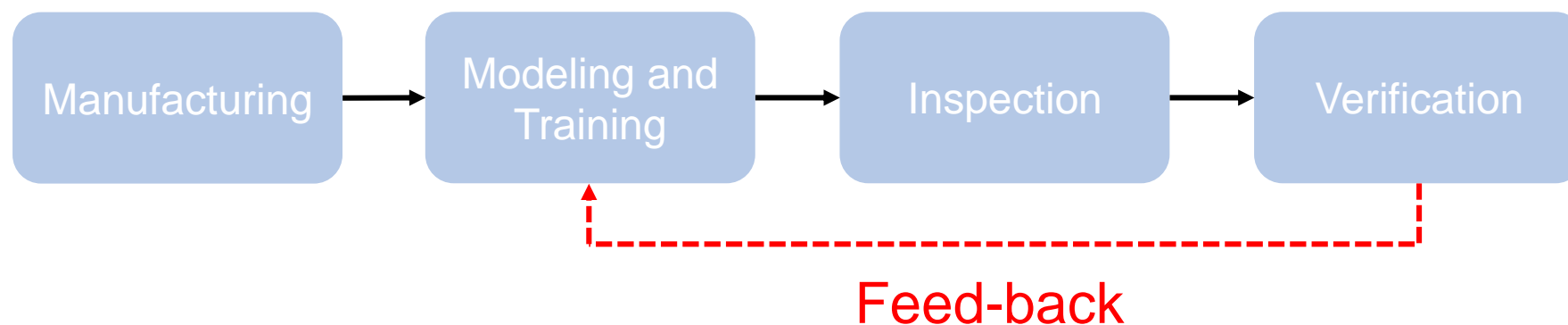
X-ray computed tomography

- High resolution 3D image of pipeline
- Good image of cracks (submillimeter resolution)
- Very expensive
- Specimen has to be cut out (destructive)



IMPROVED APPROACH

- Strong potential to improve the quality of our models
- Feeding back verification results into machine learning models
- Real defects are better than artificial defects
- A lot of distributed and non standardized data in-house



WHAT MAKES THIS FEEDBACK LOOP HARD TO ACHIEVE?

- How to get the verifications from the clients to the data scientists?
- We don't define what is verified
- Data is not clean
- Data is not aligned



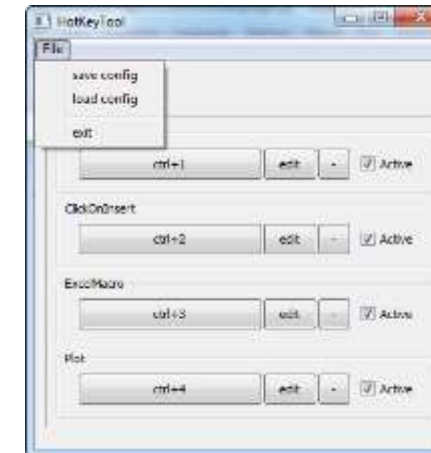
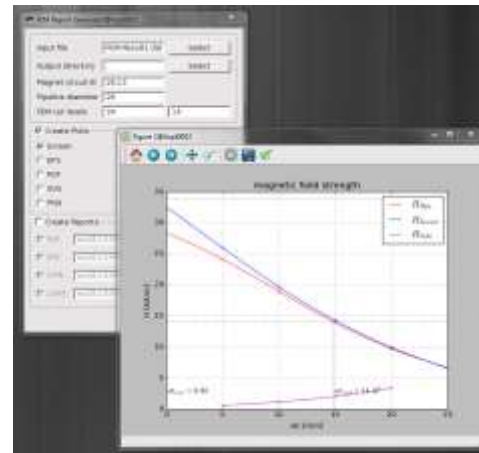
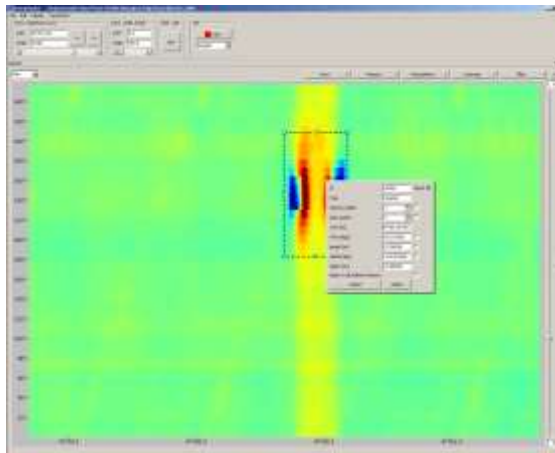
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HOW TO GET THE VERIFICATIONS FROM THE CLIENTS TO THE DATA SCIENTISTS?

- Know the delivery chain
 - Client ➡ Project Manager ➡ Data Engineer ➡ Annotator ➡ Data Scientist
- Define structures and processes
- Inform and train contact persons
- Help people who can help you
 - E.g. write small tools (in Python), which help them solve their problems
 - Automate the boring stuff



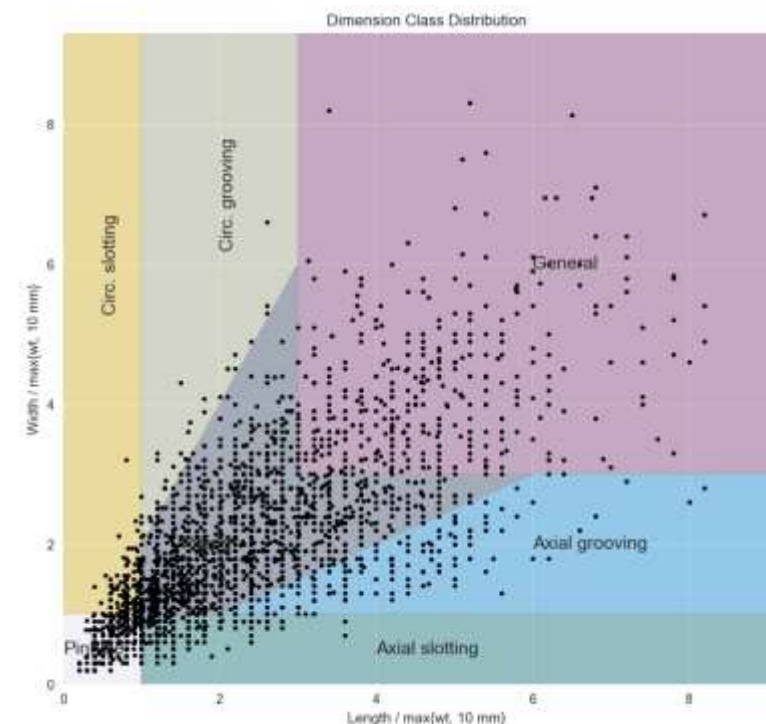
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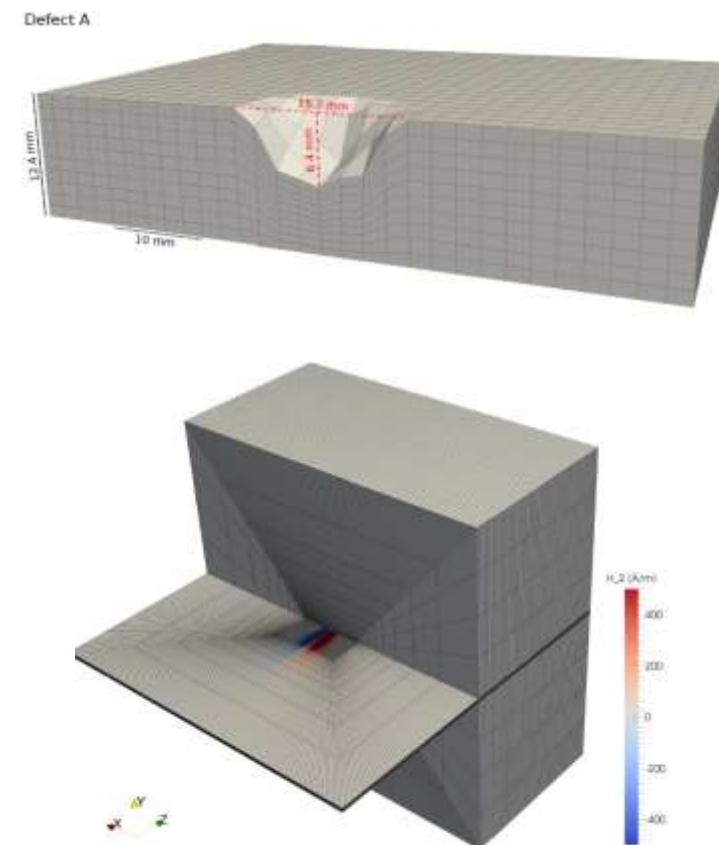
WE DON'T DEFINE WHAT IS VERIFIED

- Clients decide where excavations are done
- Very expensive!
- Fill the gaps:
 - Pull-tests and pump-tests
 - Laboratory measurements
 - Synthetic data (PyCon2017: Synthetic Data for Machine Learning Applications)



SMART WAYS TO FILL GAPS

- Synthetic defects
 - Distort laser-scans
 - Basic geometric shapes (e.g. ellipsoids)
 - Simulate corrosion growth (3D Cellular automata)
- FEM Simulations
- How to scale?
 - 15 min for each FEM simulation on one core
 - Distributed computation with a Docker cluster



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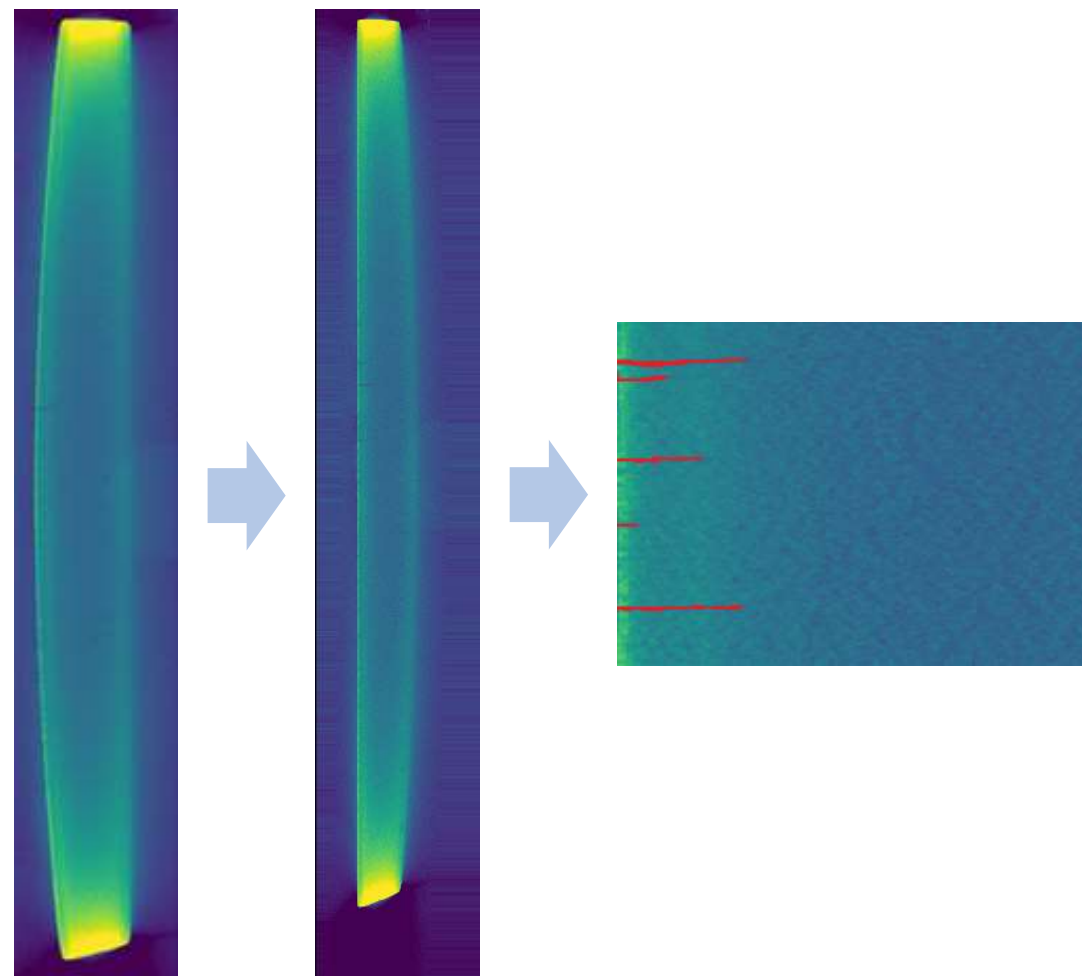
DATA STANDARDIZATION

- Clients use different data formats
- Flexible converter tools
- CSV or HDF5 as data container
- MongoDB for the meta data
- Proper interfaces for data access
- Data storage
- IT-support



DATA NORMALIZATION

- Data is not comparable to other data
 - Wall-thickness
 - Pipeline curvature
- Image processing
 - Filtering
 - Edge detection
 - Hough transformation
- Extraction of important information



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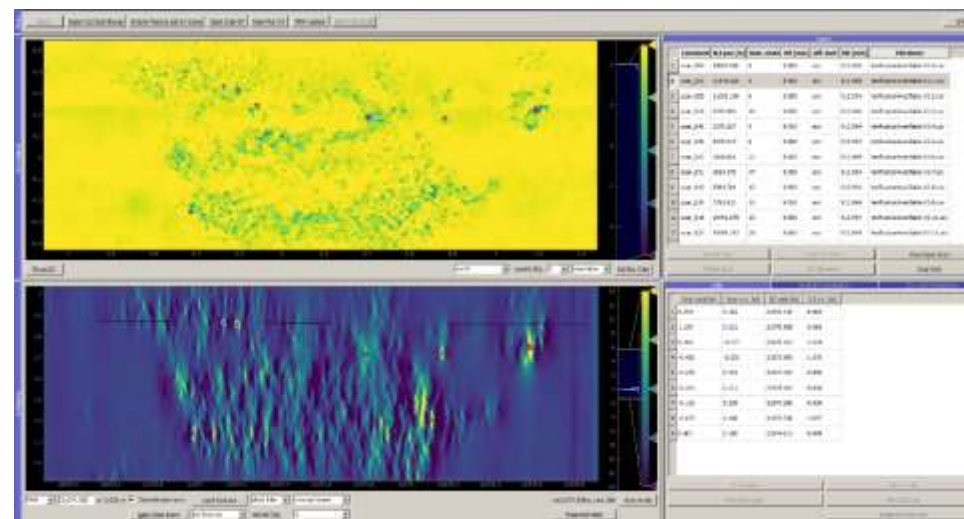
ALIGNMENT

Scan Alignment Tool

- Tool to align laser-scans and ILI-data
- Combines data from different sources
- Manual alignment

Automated alignment

- Template matching
- Direct comparison of laser-scans and ILI-data is not possible
- FEM simulations bridge this gap



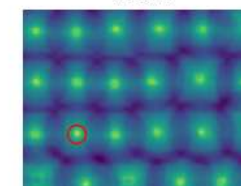
template



image



'match_template'
result

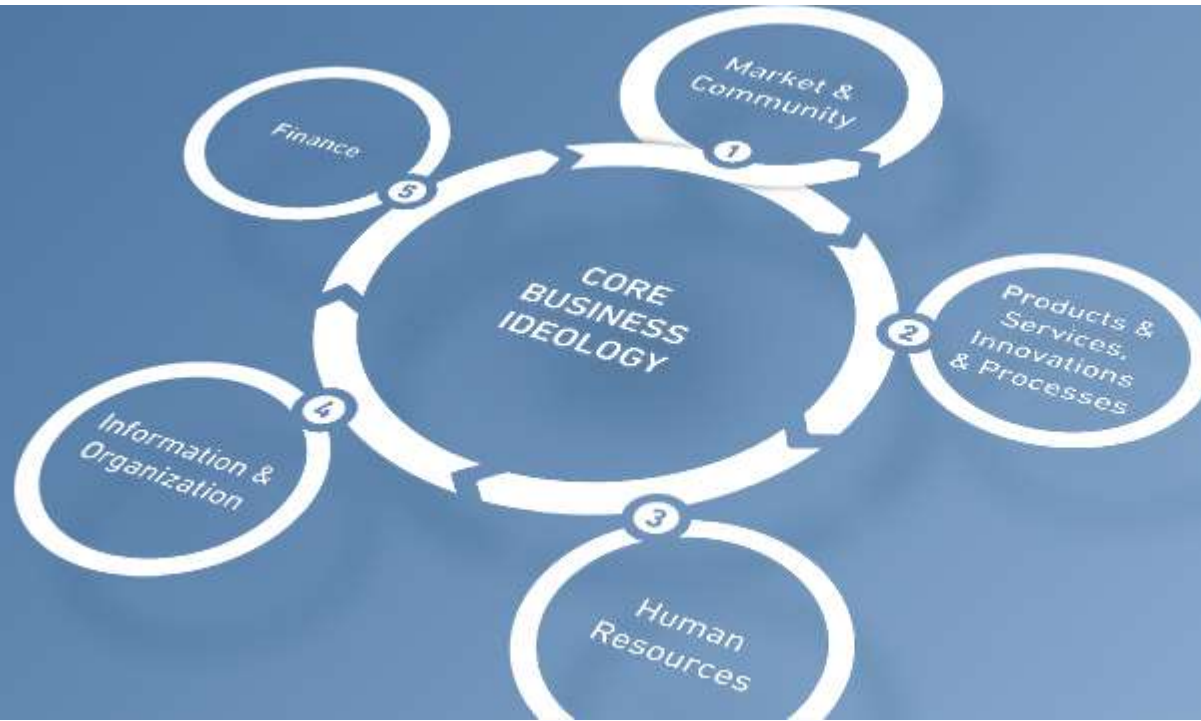


SUMMARY

- Data science challenges in ILI
- Misconception about data science
- Our classic and improved approaches
- Feeding back verification data is hard
- Various methods to tackle these challenges

Conclusion

- It is hard to feed-back validation data into our models
- But it is worth it!
 - Strong increase of classification and regression accuracies
 - Completely new approaches became possible



**THANK YOU FOR JOINING
THIS PRESENTATION.**
