

Innovating tank bottom inspections with MFL Ultra sizing and machine learning

AI/ML use case

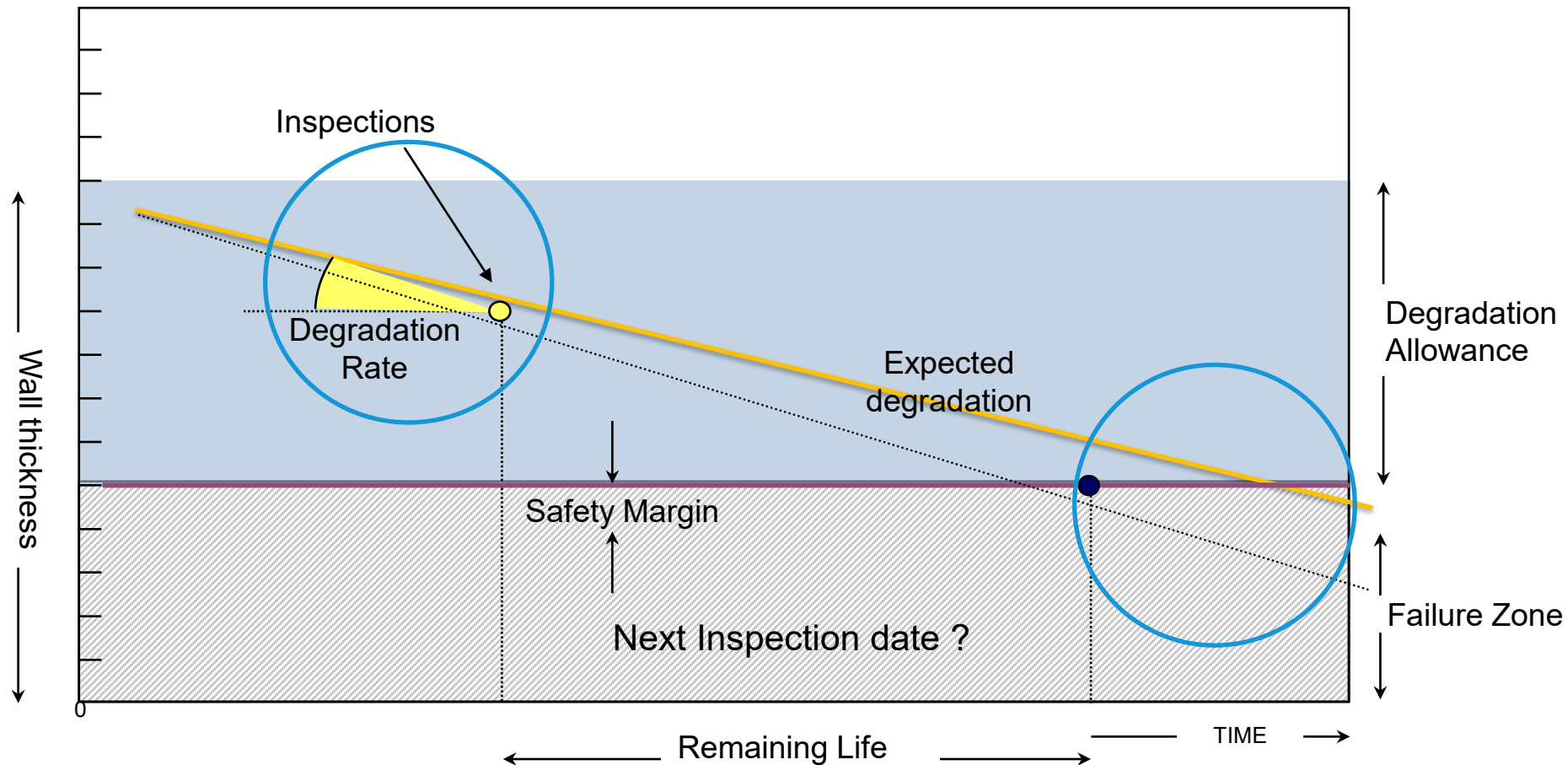
Josh Rawson | ROSEN Group | Germany | 2025

All roads lead to the bottom

- Highest corrosion risk occurs at the bottom.
- Most tank failures originate at the floor due to undetected or underestimated metal loss.
- Difficult to monitor externally.
- Primary input to API 653 calculations – remaining wall thickness and corrosion rate from the bottom drive the Next Inspection Date (NID).
- Risk-Based-Inspection (RBI) hinges on bottom data – more accurate detection and sizing allow safe extension of intervals and optimized maintenance planning.



Extending inspection intervals through RBI process



More accurate measurements lead to:

- Extended asset lifetime
- Lower probability of failure reducing risk



RBI Optimization

Industry problems & our response

Problem #1

Early Detection of Corrosion
& Degradation



Solution #1

Reporting Threshold shall be reduced
from 20% to 10%.
(e.g. in pipelines – 5%).

Problem #2

Optimizing Maintenance
Schedules



Solution #2

Next Inspection Interval – shifting from
fixed intervals to calculated, risk-based
planning.

Problem #3

Improving Safety



Solution #3

Improved POD and accuracy without the
need for UT check-up's (i.e. reduce time
for personnel in confined space).

Problem #4

Improving Cost Efficiency



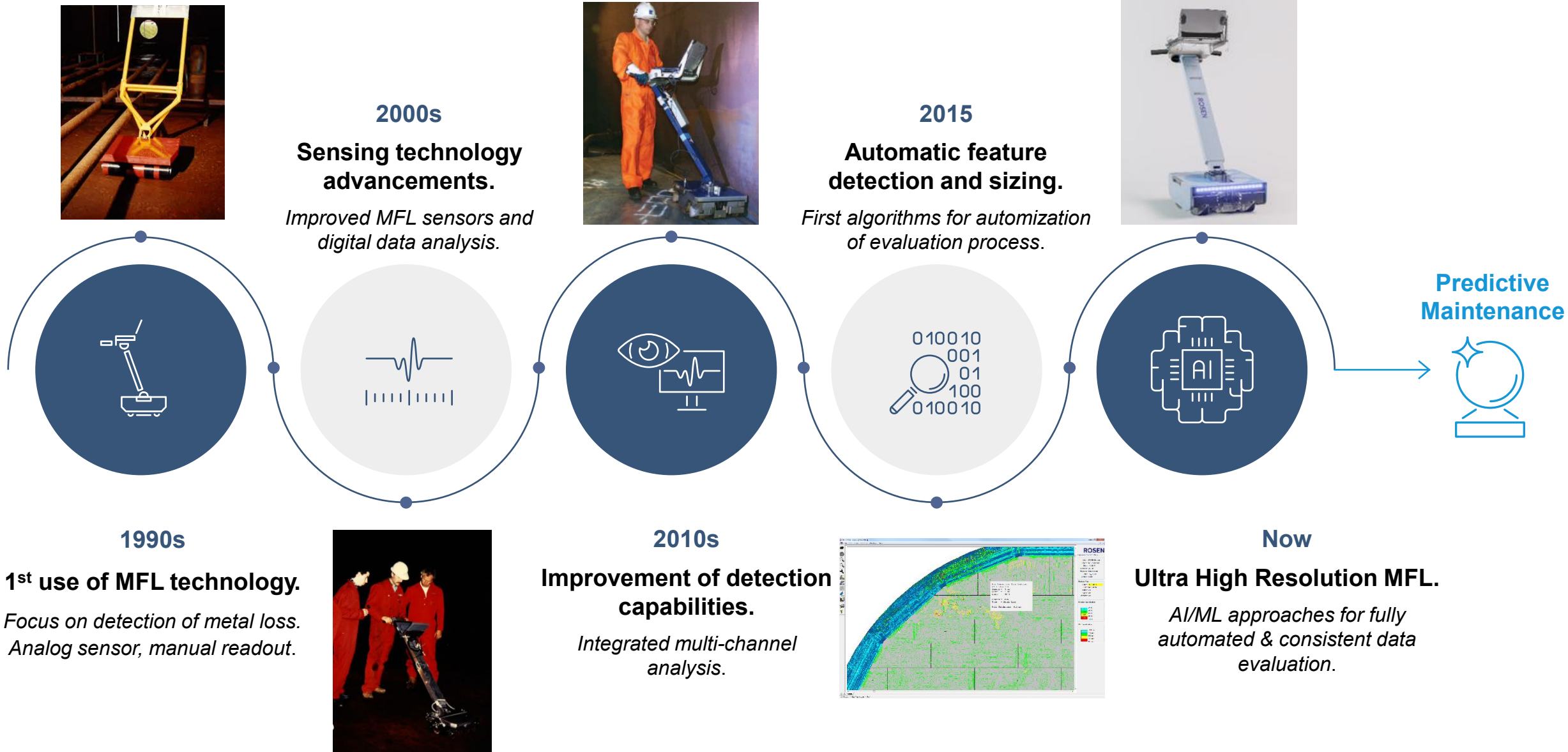
Solution #3

Repair Criteria – calculated, based on
actual degradation rates.
Fast turnaround – feature list available
when leaving the tank after scanning.

The evolution of MFL in tank bottom inspection

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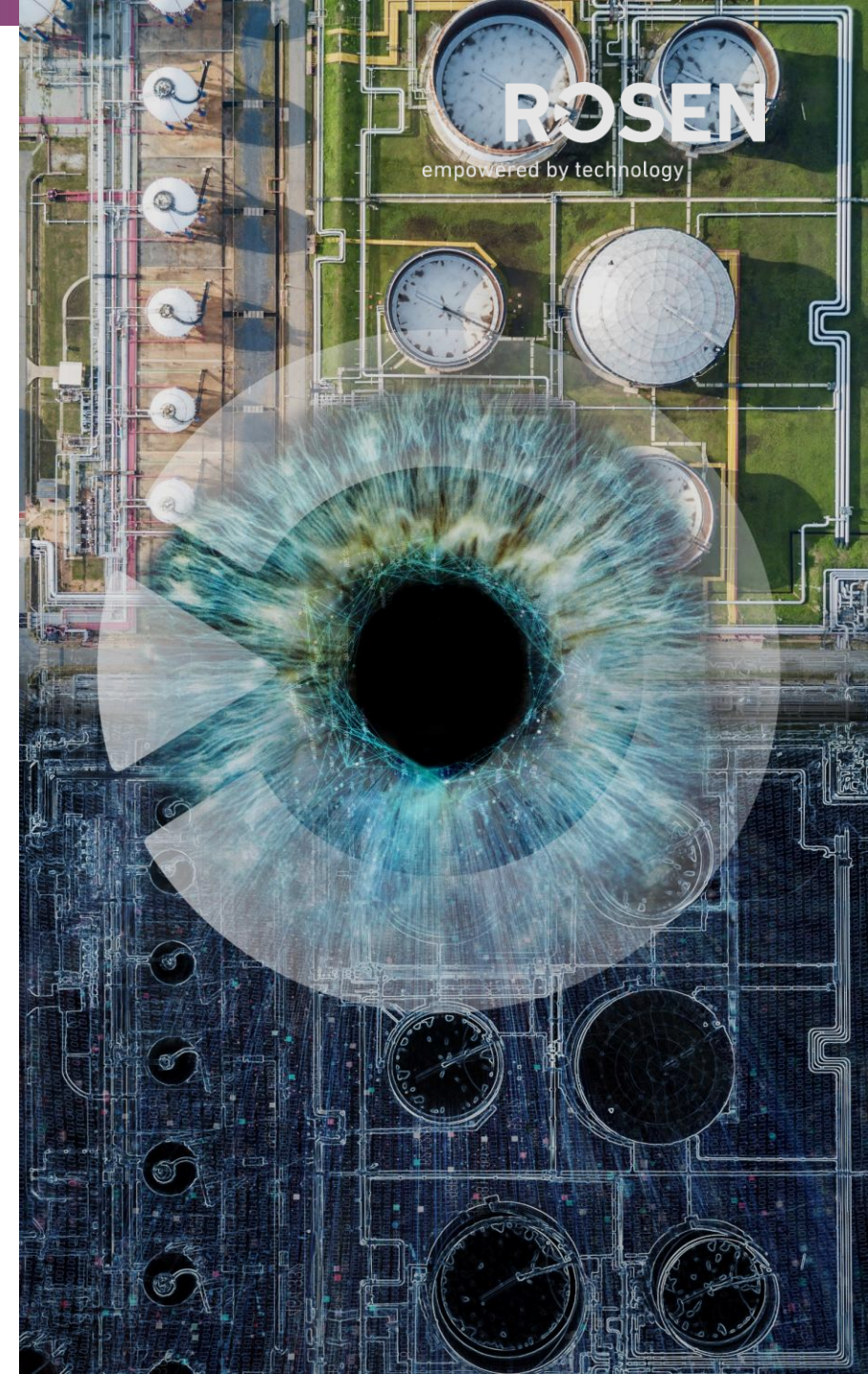
empowered by technology



Ultra high resolution MFL Data

What is the smallest defect we could find?

- **Data Quality** - Reliability of results by far exceeding conventional MFL or EC scanners
- Indications $\varnothing > 2\text{mm}$ are detected
- Indications **as small as 10% ML (0.1t)** are detected (e.g. for a plate of 6.35 mm \rightarrow 0.6mm or .25" \rightarrow .023")
- Sizing **accuracy is +/-4%**
(e.g. for a plate of 6.35 mm \rightarrow +/- 0.3mm or 0.25" \rightarrow +/- .011")
- **Inspection turnaround time** is faster due to automated feature detection and sizing
- Coatings up to **6 mm/0.24"** do not need to be removed



Extending inspection intervals

Maximizing inspection intervals through best-in-class sensitivity

- New tank bottoms are commonly inspected 10 years after commissioning
- Normally **only limited corrosion is present**
- **Next Inspection Dates (NID)** derived from the inspection **often do not reach the allowable limits** given by the regulator (new tanks as well as older tanks)
- **Extremely low detection & repair thresholds** are needed in order to actually achieve a calculated inspection interval of 20 years or greater (according API 653):

Example: Floorscanner “A”

- Tank bottom constructed 2010
- Bottom plates thickness : 0.250 in. / 6.35mm
- Inspection in 2020 (10 years after commissioning)
- Detection: **20%**-100% features
- Repair threshold: **>20%**
- Next Inspection: **10 years**

Example: Floorscanner “B”

- Tank bottom constructed 2010
- Bottom plates thickness : 0.250 in. / 6.35mm
- Inspection in 2020 (10 years after commissioning)
- Detection threshold: **10%**-100%
- Repair threshold: **>10%**
- Next Inspection: **25 years** (maximized to 20 according API 653)

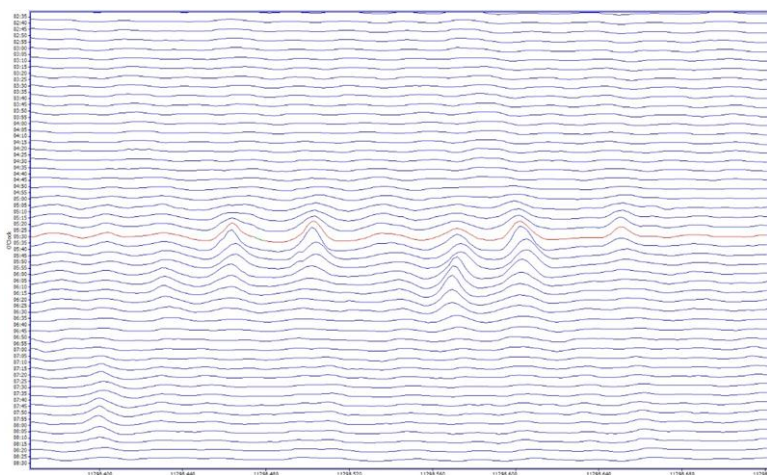
Complexities in Ultra High-Resolution MFL-technology

Highlighting the complexities that arise alongside our ultra high-resolution inspection advancements.

Generation of Ultra High Resolution MFL data

154

sensor channels



Traditional MFL resolution

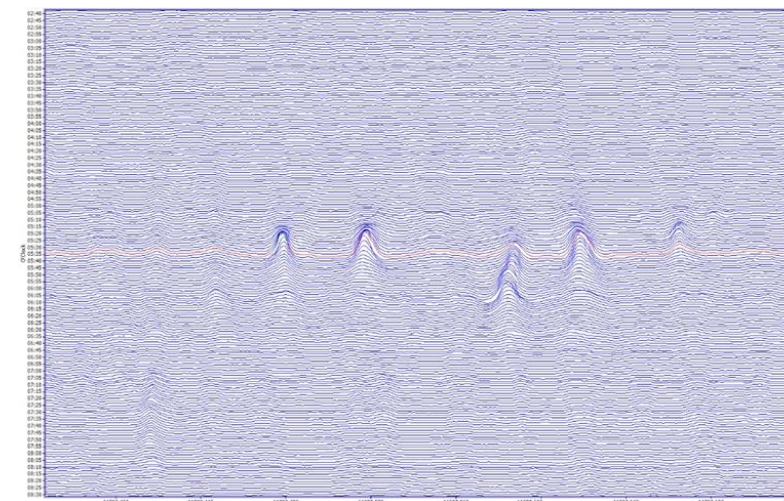
6.7× higher
resolution



Increased
**sensitivity &
precision**

1029

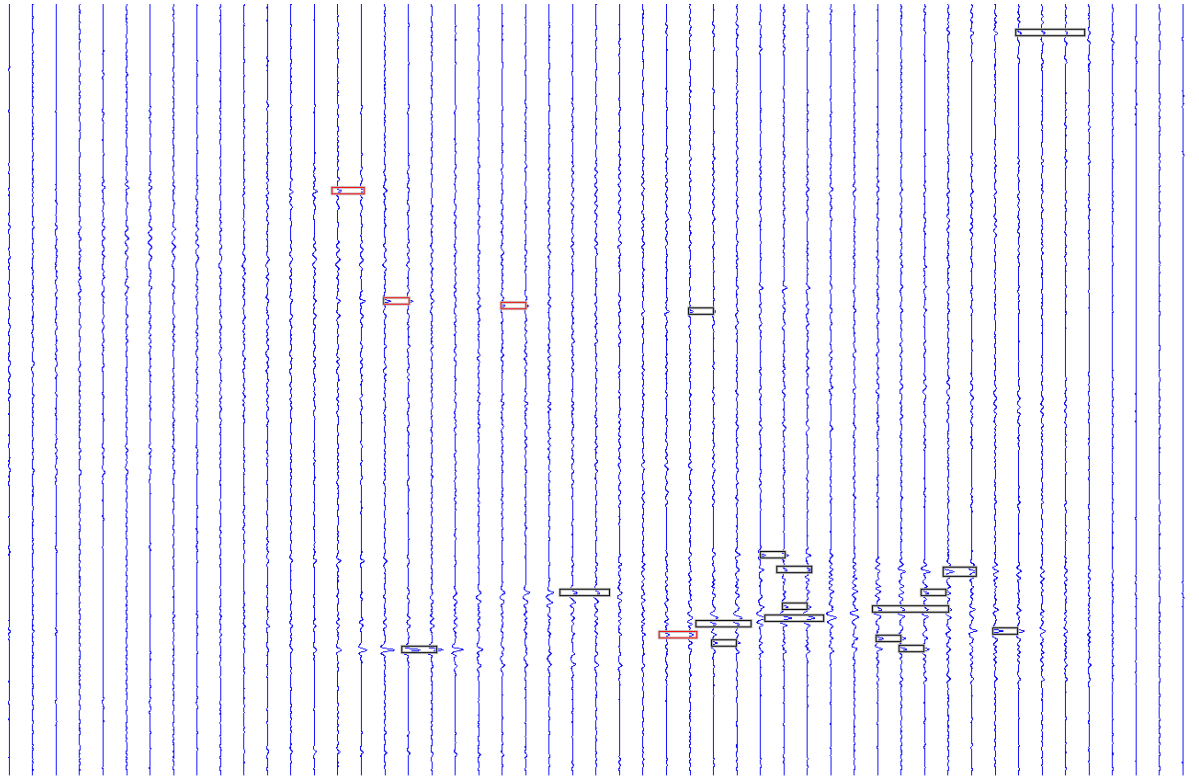
sensor channels



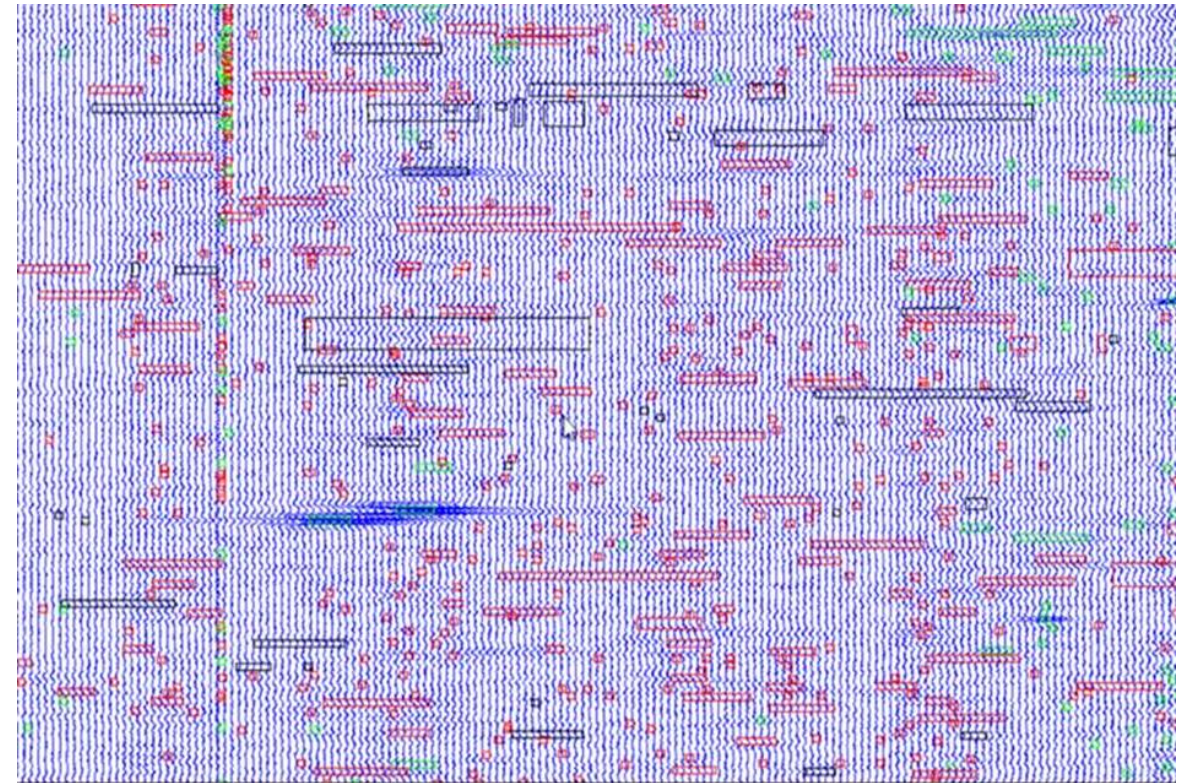
Ultra High Resolution

Complexities introduced by Ultra High Resolution

Increased number of false calls



Traditional MFL resolution

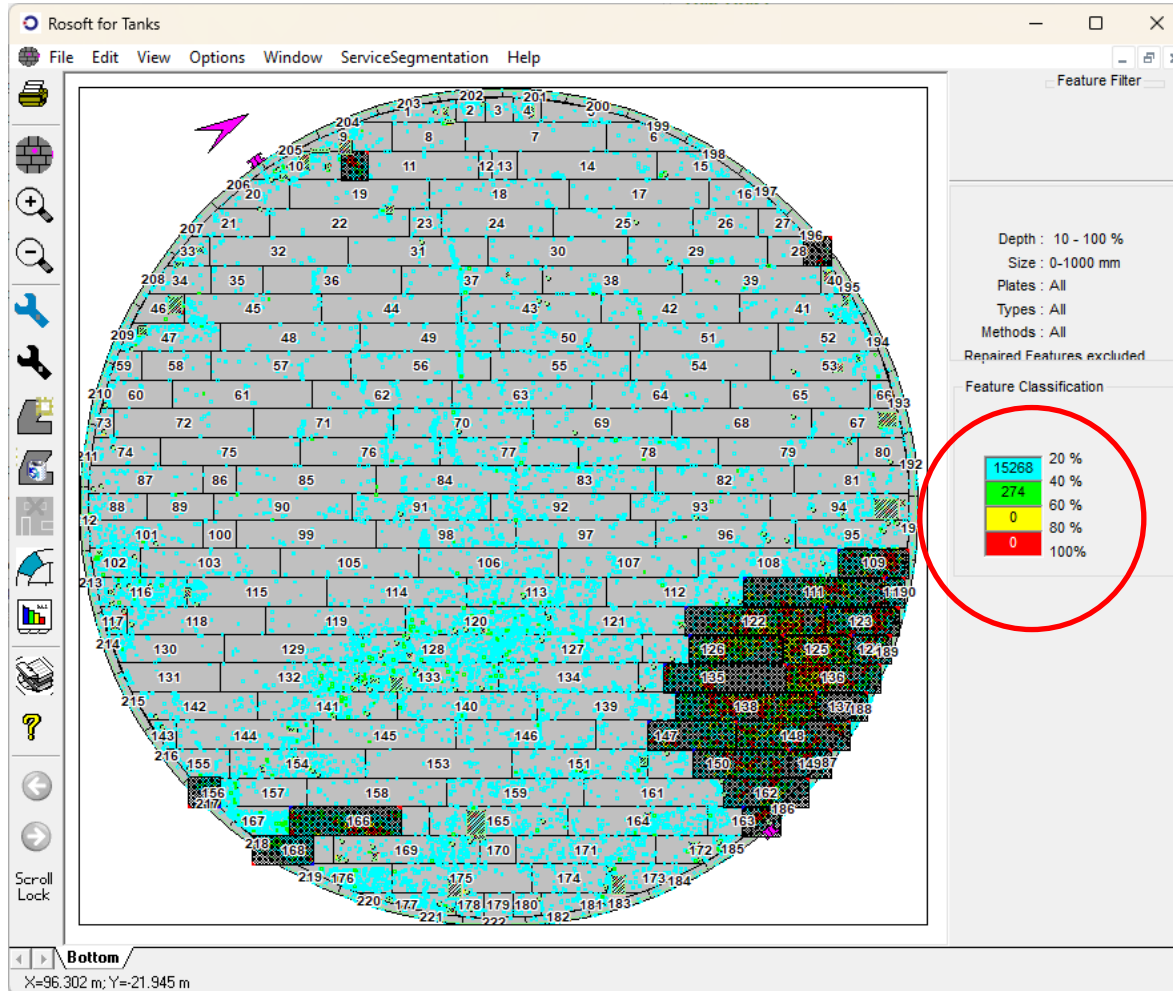


Ultra High Resolution

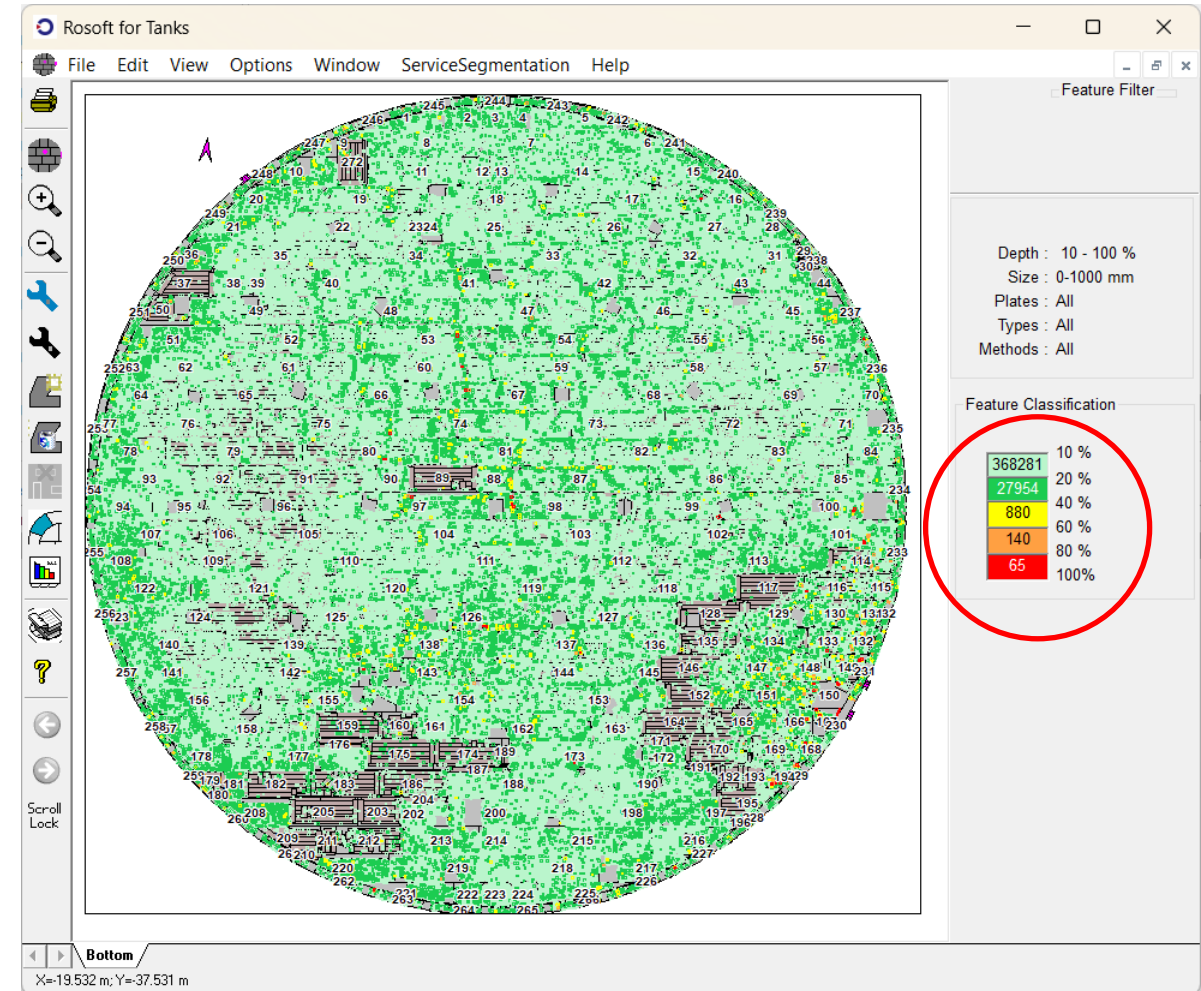
Complexities introduced by Ultra High Resolution

Increased number of indications to be assessed

Inspection results with **traditional** resolution (2019)



Inspection results with **Ultra** resolution (2024)



Complexities introduced by Ultra High Resolution

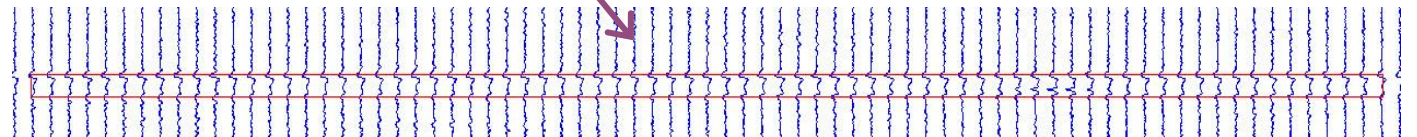
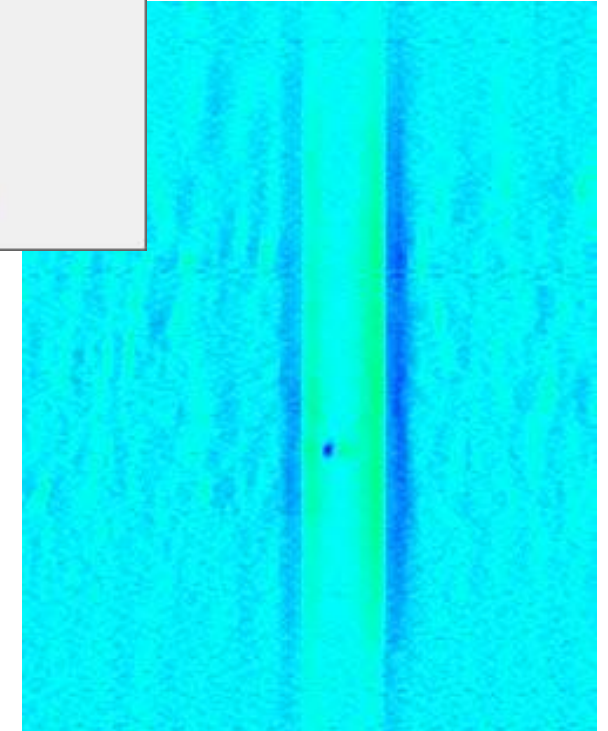
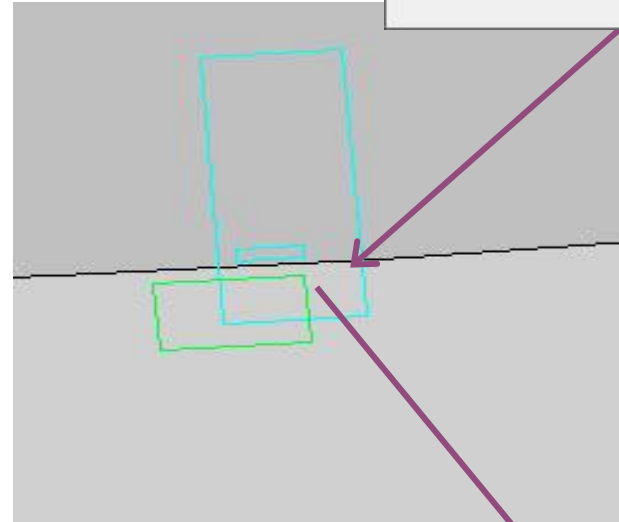
False call example - Tack welds

Tack weld, often left over by welders protrude from the bottom plate usually found near other bottom plate welds.

Tack welds in line plot usually resembles the letter **M**

Inspection Side Feature:
No.: 50.0.156
Tool ID : T56
length = 54 mm
width = 133 mm
depth = 21 %

Method = TBit
Type = Metal Loss

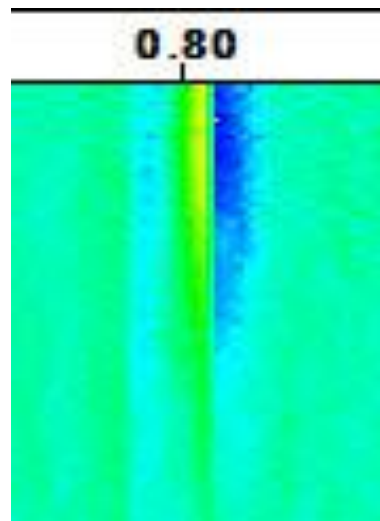


Complexities introduced by Ultra High Resolution

False call example - Settlement

Settlement indication usually appears as a straight line that runs along the width of the line plot.

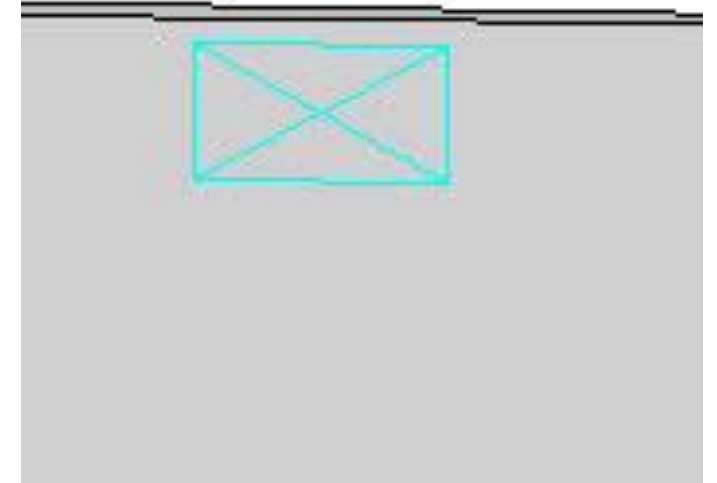
C-Scan



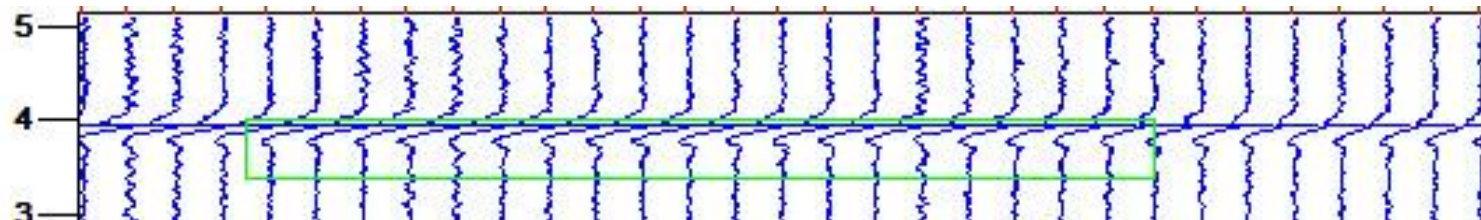
Inspection Side Feature: (disabled)
No.: 51.0.58
Tool ID : T56
length = 65 mm
width = 35 mm
depth = 27 %

Method = TBit
Type = Metal Loss

Plate
View



Line plot

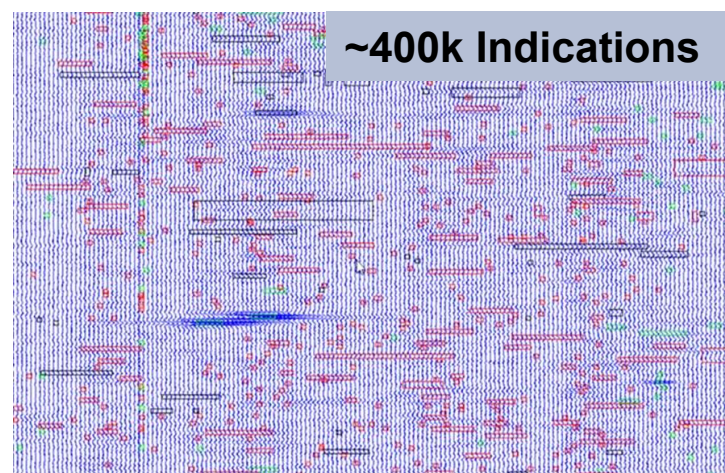
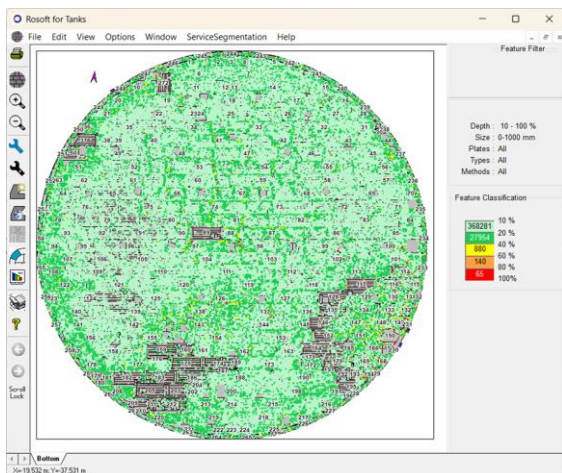
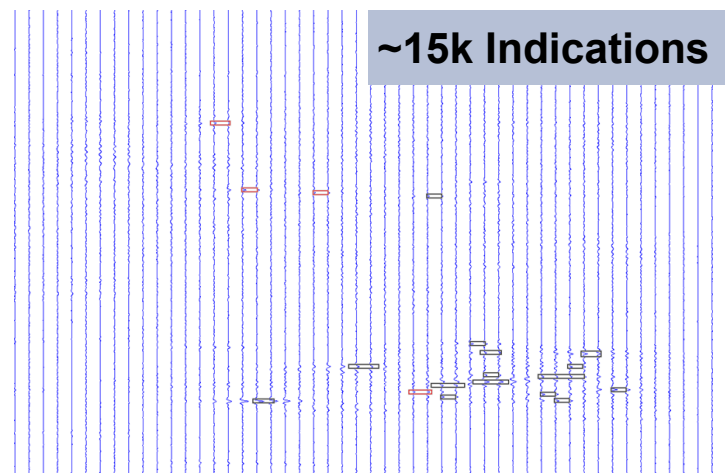
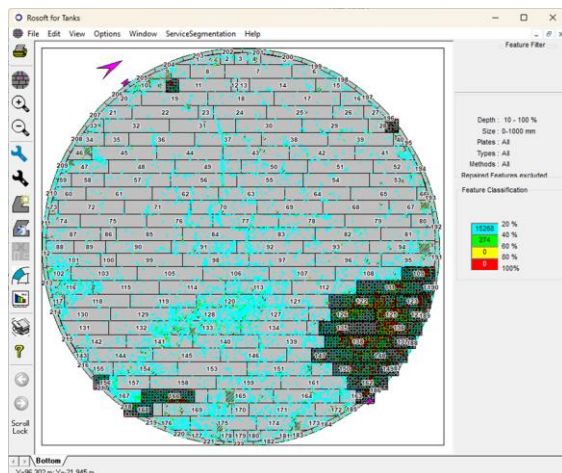


The solution - AI/ML supported automation

Huge amounts of detected indications cannot be assessed manually anymore. That's why AI and ML approaches need to support and efficient and effective data evaluation process.

Quality and time is key

The importance of automation in data evaluation



Example TBIT Ultra inspection:

Tank Ø: **60m**

Inspection duration: **3 days**

Defects found in Standard mode: **15.542**

Defects found in Extended mode: **397.320**

Full auto online feature sizing: **0 seconds**

Other systems:

Detection + marking: ? days

Stop on defect. Additional time: 15.000×3 minutes =

750 hours additional = **75 man-days**

Is this time (cost) taken?
What if it isn't?

The solution

Combination of automatic & manual data evaluation

Higher
sensitivity

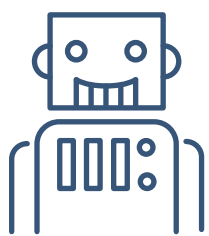


Higher false
call rate

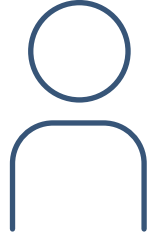


**How to filter
between true
and false call?**

Two stage data evaluation:



1. Computer
(Algorithms)



2. Evaluator
(Human)



- Fast
- Quality fixed and equal across the globe
- No human error effect on the majority of data
- Can cope with unusual data



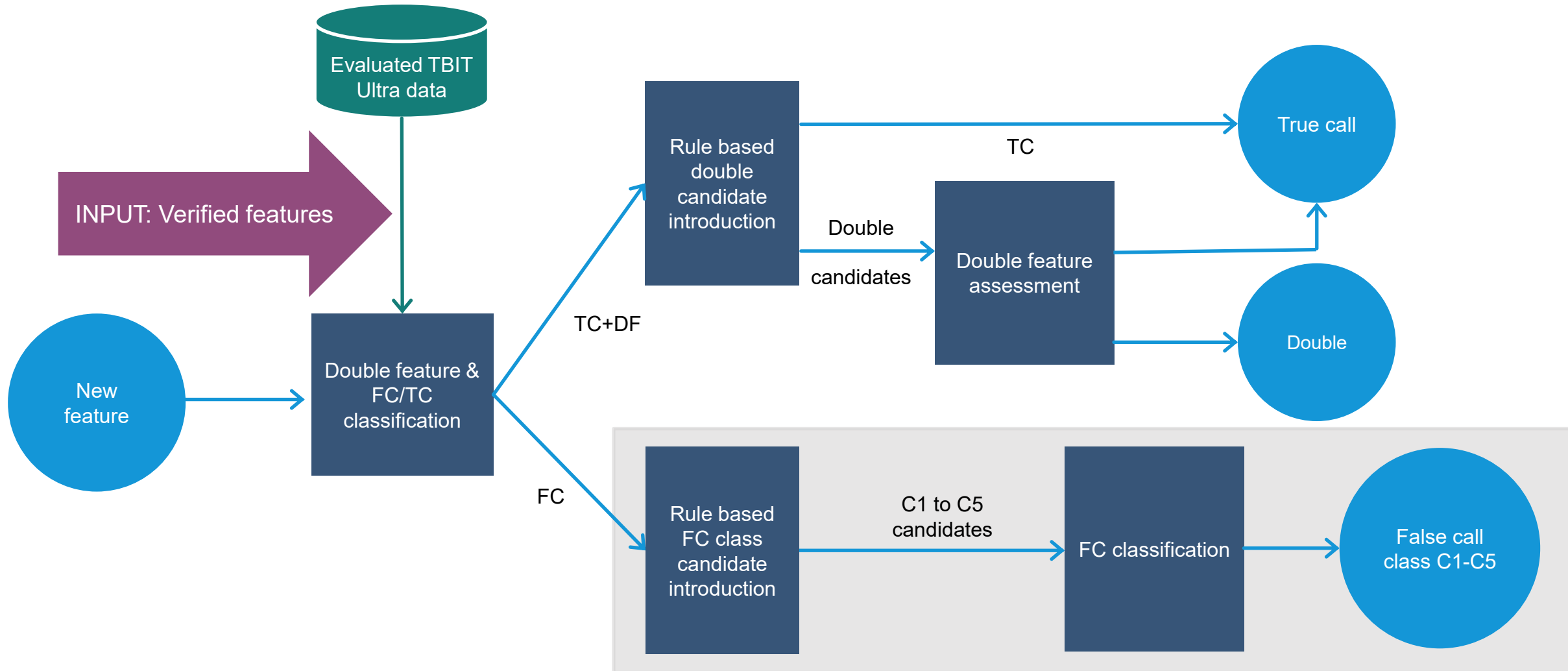
85% of the evaluation
work is taken over by
the algorithm

3% of the evaluation
work is still performed
by the Evaluator

12% true indications

The solution

Algorithm for automatic double feature & false call elimination



Conclusion

How is Ultra High resolution MFL data revolutionizing the tank bottom inspection approach

Summary - Benefits of Ultra High Resolution MFL data



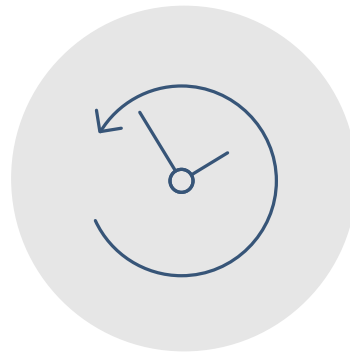
Sensitivity

Highest sensitivity allows for early detection of small features



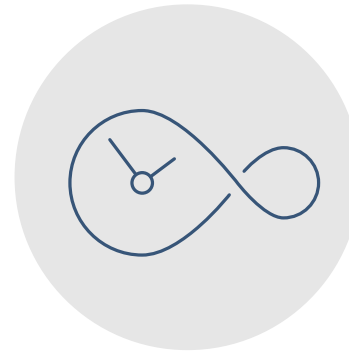
Confidence

Confidence and reliability in MFL data analysis



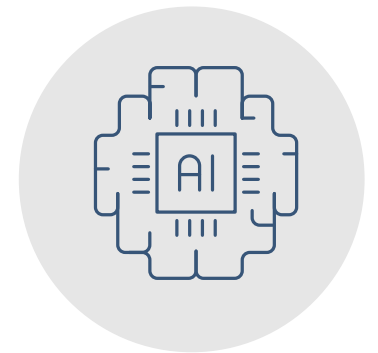
Time

Reduction of hours due to automated feature detection and sizing



RBI

Results support RBI assessments, optimizing inspection intervals and NID determination



AI/ML

Enabler for AI-powered predictions

"The Big Data revolution is less about collecting more and more data. It is about collecting **the right data.**"

— Seth Stephens-Davidowitz, *American data scientist & economist*



The logo for ROSEN features the word "ROSEN" in a bold, white, sans-serif typeface. The letter "O" is stylized as a circle with a horizontal line passing through its center, resembling a play button or a film reel. The background is a dark blue gradient with a thin horizontal bar at the top transitioning from teal to purple.

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