

Appendix 9

Current topics relevant to CUI

Dave Harvey (TWI)

EFC-NACE Italia Section Joint
Meeting, Venezia 31 March 2006

Current Topics Relevant to CUI

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CUI – Carbon & Low alloy Steels

- Localised wastage of areas in contact with water, held by insulation or between steel and insulation
- All insulated equipment operating between -5°C and 200°C at risk
- Highest corrosion rates between $60-120^{\circ}\text{C}$
- 1.5 mm/y typical corrosion rate; but 3 mm/y reported
- All thermally insulated surfaces operating between -5°C to 200°C and service life $> 10\text{yr}$ should receive adequate corrosion protection before application of thermal insulation



CUI – Austenitic / Duplex Stainless Steels

- Prone to high rates of localised attack and stress corrosion cracking under certain conditions
- Critical temperature, Cl⁻ level, pH
- Pitting threshold temperature e.g.
 - 304L >25 °C; 316L >36 °C; 2204 duplex >90 °C
- Stress corrosion cracking threshold T e.g.
 - 304L/316L >50-60 °C, 25Cr SDSS >100 °C
- Corrosion protection required on process equipment operating above threshold T:
 - Austenitic and low Mo DSS >25°C
 - DSS, SDSS and super austenitic SS >80°C

CUI – Protective Coatings

- Effective method of delaying onset of CUI:
 - Thermally sprayed Al has potential to provide protection for full service life (> 20 years)
 - Viable option for new build
 - Same surface prep QC required for TSA / painting to that for non-insulated surfaces
 - Al and SS foils are alternatives
- New build best opportunity to minimise risk of CUI:
 - Process equipment external condition and access most favourable
 - For many cases - only time steel substrate is accessible when at ambient temperature

CUI Prevention Field Application: Comparison of TSA & Paint

Features	TSA	Conventional Paint
CUI Protection	25 to 30 yrs; maintenance-free; inspection-free	5 to 13 yrs; tends to low side for on-line application
Protection in cyclic service	Yes	No effective paint system
Upper continuous Operating Temperature	480°C	175°C
Schedule impact	None - one coat application	24 hrs typically; multi coats required
Environmental impact	None	Must meet VOC & disposal regulations
In-Place cost ratio	1,05 to 1,20	1,0
Durability	Very resistance to mechanical abuse. Minor damage does not result in CUI	Very susceptible to mechanical abuse. Any damage results in CUI
Required surface preparation	White/near white (SA 2.5)	White/near white (SA 2.5)
Application method(s)	Twin Arc spray or flame spray	Spray, brush and roller
Application accessibility	Arc/spray head to within 30° normal to surface	Brush/roll restricted access but life decreases
Application temperature limit	None but service must be dry	Ambient to about 60°C
Work Permit required	Hot work	Hot work

TSA / CUI Joint Industry Project

- PR9483: Prevention of corrosion under insulation of steel with TSA (launch May 2005)
- Objective:
 - Demonstrate long-term mitigation of CUI using TSA
- Work scope (provisional)
 - C-steel substrate
 - Insulation: Rockwool, expanded Perlite, closed cell foam
 - Benchmarking vs unpainted, painted, Al foil
 - Constant T: 40°, 90° & 140°C
 - Cyclic T: -5° to +150 °C
 - Aggressive electrolyte (3.5% NaCl) & de-ionised

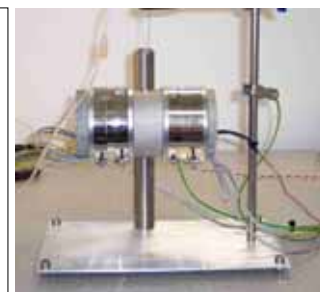
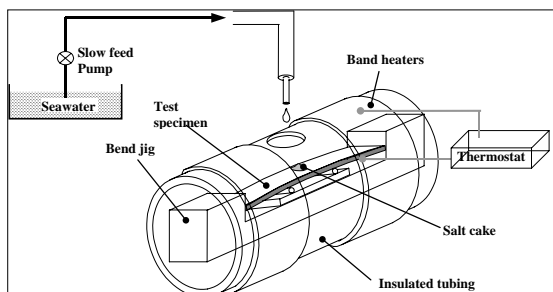
Susceptibility Score Table for CS/Low Alloy Steel

Operating Temperature	Coating Status when new or last applied	Cladding/ Insulation Condition	Insulation type	Remnant corrosion allowance	External coil/steam tracing	External environment
Constantly >175°C or <-5°C	Full QA coating ≤ 8 years, or TSA* ≤ 15 years	Good to Engineering Standards or renewed (<5 years)	Contact free insulation, with regular inspection (every 5 years)	> 4mm	Not present	Inside building, not steam traced and not sweating DEFAULT NEGLIGIBLE RISK
150 - 175°C	Full QA coating 8-15 years or conventional coating ≤ 8 years or TSA* 15-20 years	Average condition, overall high integrity design & construction	Expanded Perlite, Foamglass, Closed-Cell Foam (good type)	2-4 mm	High integrity design	Low wetting rate (<20% of the time)
-5 - 49°C and 111 - 149°C	Conventional coating 8-15 years or TSA* > 20 years	Average condition, conventional design & construction	Fiberglass, Asbestos, Regular Perlite, Mineral/Rock Wool (<10ppm Cl)	1-2 mm	Medium integrity design	Medium wetting rate (20 - 50% of the time)
50° - 110°C or sweating conditions	Full QA or conventional coating > 15 years or unpainted or unknown	Poor condition, damaged/ wet/broken seals	Cal Si, Rockwool (no spec), unknown	< 1mm	Low integrity design or leaking	High wetting rate (>50% of the time) (e.g. cooling tower/deluge systems)

External Chloride Stress Corrosion Cracking (SCC) Tests

Top right - un-coated coupon after 14 days

Below - drop evaporation test rig



For further information
contact:

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Appendix 10

Recent advancements of the EFC WP15

CUI Guideline

Maarten Lorenz



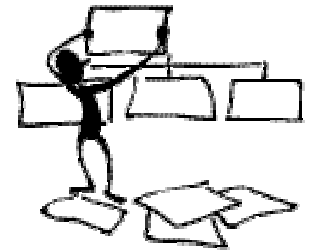
Recent advancements of the EFC WP15 CUI Guideline

Working towards a unified risk-based
approach to CUI management

EFC WP15 Meeting
Venezia, 31 March 2006

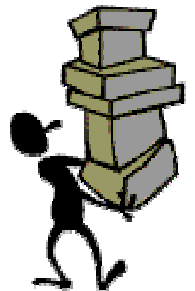
Set-up of the EFC WP15 CUI Guideline (1)

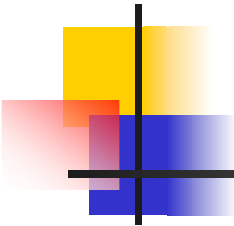
- Problem definition – cost analysis
- Ownership & responsibility – staff involvement
- Risk-based inspection methodology for CUI
 - unit level prioritisation
 - challenging the need for insulation
 - reality check
 - determination of risk of CUI failure
- Risk-based inspection & maintenance plans
 - inspection efforts
 - adequate coating and insulation applications



Set-up of the EFC WP15 CUI Guideline (2)

- NDT techniques for CUI
- Susceptible locations for CUI
- Appendices:
 - detailed information on available NDT methods
 - examples of CUI
 - best practices from the field
 - designing out CUI

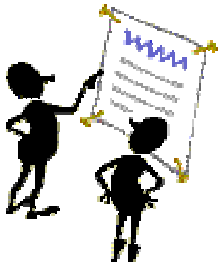




Objective of the EFC WP15 CUI Guideline

To provide:

- A high level risk-based approach
- Adequate inspection & maintenance strategies
- Best practice from the field



...in order to manage CUI effectively

NOT to provide:

- A detailed prescription of when to inspect for CUI
- Approved NDT techniques
- Mandatory maintenance methods





Use of the EFC WP15 CUI Guideline

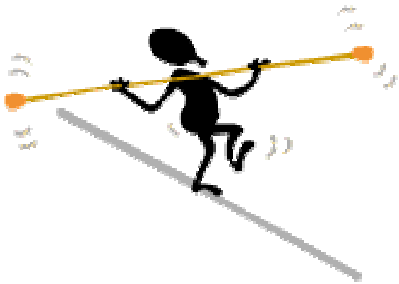
- To design optimised plant-specific CUI inspection & maintenance plans
 - tuned to local conditions
- To facilitate budget approval for implementing effective CUI management plans
 - life cycle cost philosophy
- To share experience, best practices & state of the art
 - living document





Way forward from now...

- Finalising editing of draft document
- Include representative examples
=> your contribution is appreciated !
- Aiming to issue first version mid 2006



**CUI cannot be eliminated,
but it can be managed!**

The EFC WP15 CUI Guideline
intends to contribute to reducing unexpected
CUI failures, hence improving safety and availability



Recent advancements
of the EFC WP15 CUI Guideline

Thank you for your attention...
...any questions?

Examples...(1)

CUI of 4" insulated gas compressor recycle line



Examples...(2)

Random CUI on insulated storage tank



Examples...(3)

CUI of insulated small bore connections



Examples...(4)

Very local CUI on insulated painted pipelines



Appendix 11

RBI tank inspection strategy

Algra Rindert (RTD Group)



RTD Tank Inspection Strategy



Purpose

Safeguarding the integrity of the tank:

- while keeping the tank in service as long as possible
- within safety, health and environmental framework

Maximizing tank availability:

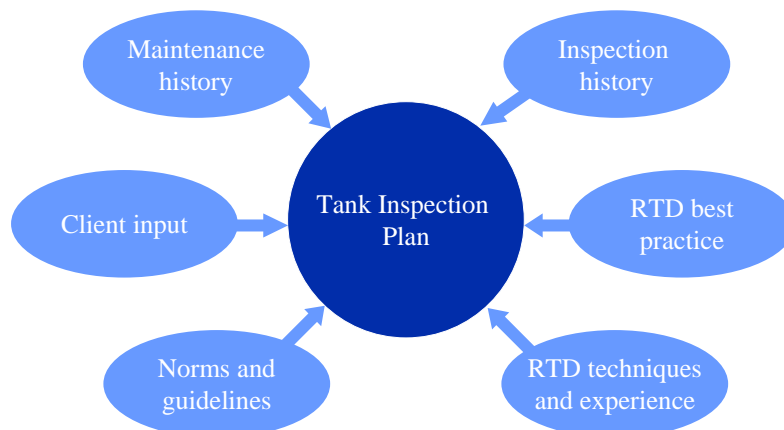
- through longer inspection terms
- by smartly combining planning between maintenance and production departments

Scope of work

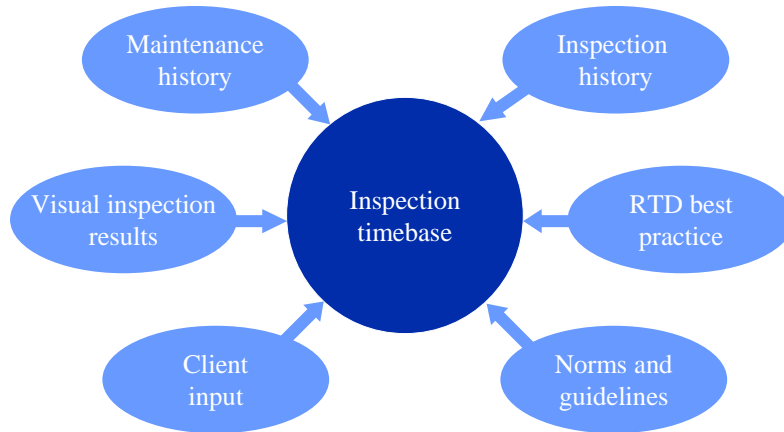
RBI and inspection plans answer the following questions:

- What to inspect? (what objects, what part of the objects)
- When to inspect? (inspection planning)
- How to inspect? (what techniques, what procedures, to what extent)

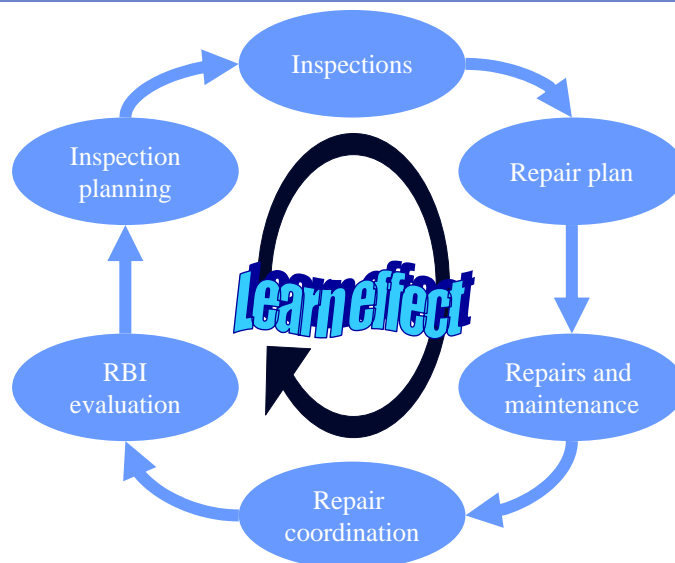
Tank Inspection plan



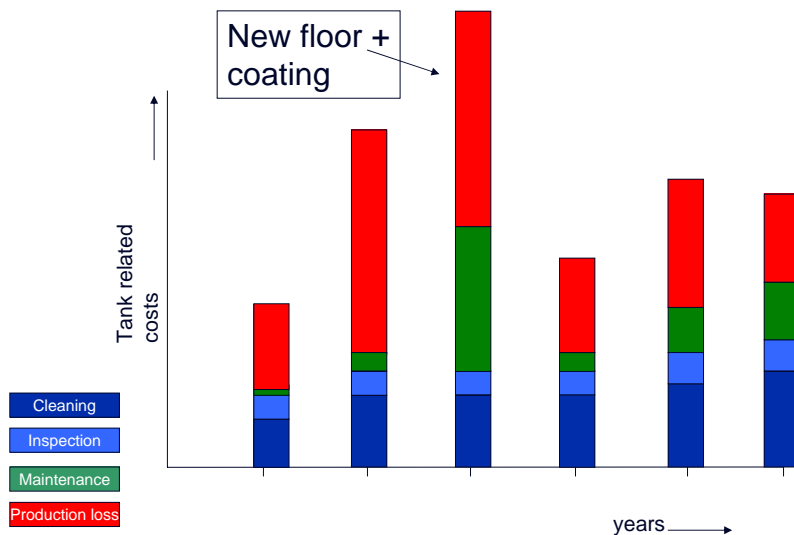
Inspection timebase



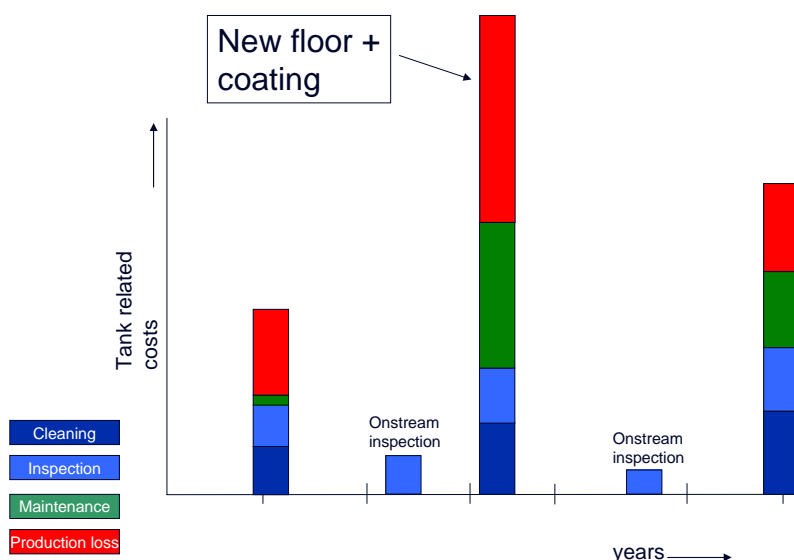
Inspection cycle in closed-loop



Tank 53 customer X



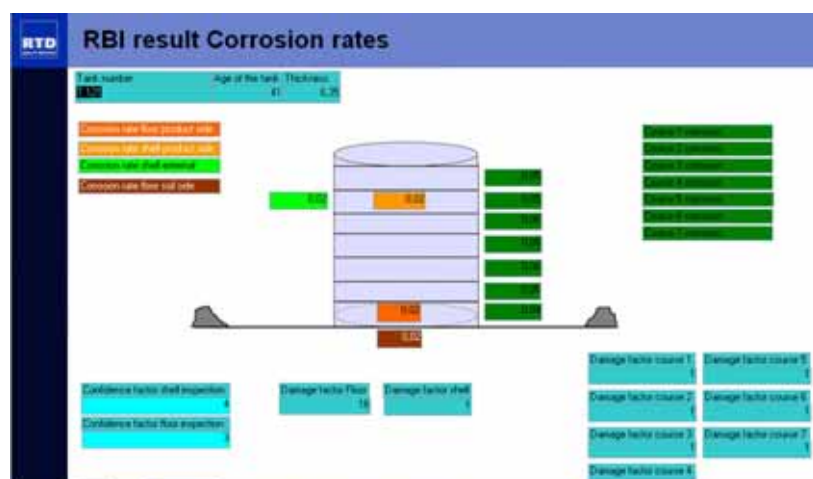
Tank 53 customer X optimal inspection cycle



API 581 RBI

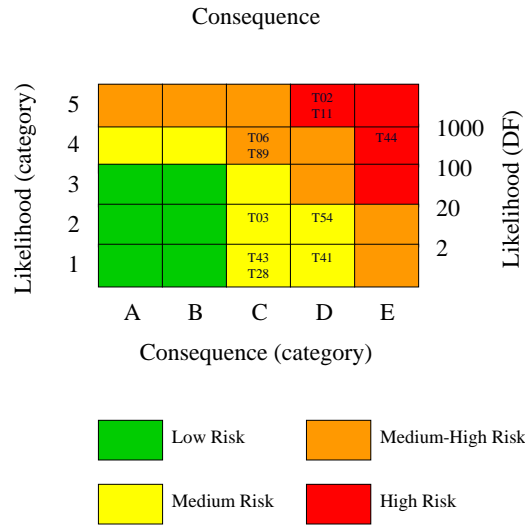
- Internationally accepted guideline
- Based on an extensive statistical study
- Very flexible through safety factors
- Inter-linked with API 653 and inspections
- RTD has written tailor-made software
- Advises on what to inspect, when to inspect and how to inspect (quantify and plan NDT!)

RBI software made by RTD

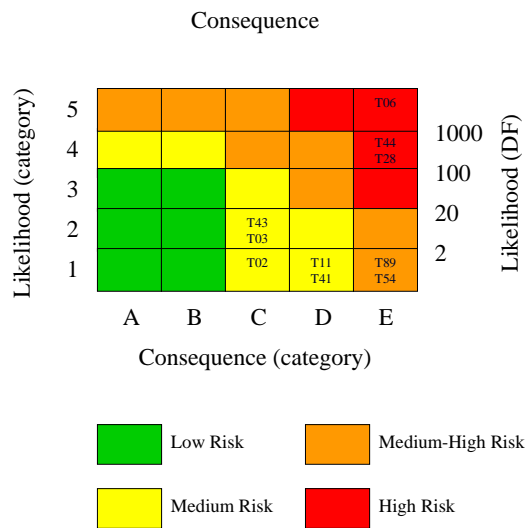


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Telephone: +31 (0)10-2082208 - Fax: +31 (0)10-4159022 - E-mail: rtd@td-group.com - Internet: www.rtd-group.com

Client case: risk results floor



Client case: risk results shell



Client case Tank T44 (inspection)

What if we keep inspecting tank T44 using API 581 and API 653?

Present situation:

- Damage factor floor is 145
- Damage factor shell is 1
- Corrosion rate floor is 2.1 mils per year
- $ar/t = 0.43$
- Inspection confidence is 0,7
- Inspection effectiveness is D1
- Next internal inspection is advised in 2005

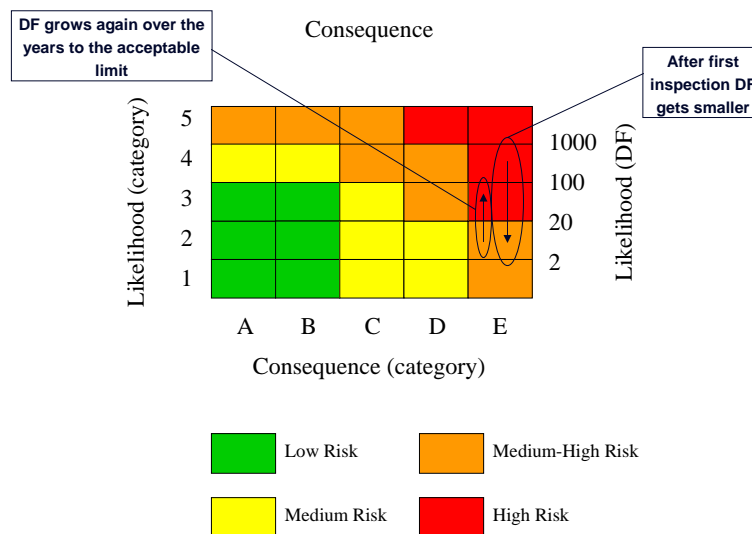
Next internal inspection (2005)

- Tankfloor is scanned, a repair plan is made and some patch plates are installed
- UT measurements are taken for the floor and corrosion rate is indeed 2.1 mils per year
- Visual inspection is performed
- Only T-scan on shell for corrosion pits, it is still in a good state

API 581 re-evaluation tankfloor (inspection)

- Consequence of failure stays the same
- Inspection confidence goes from 0.7 to 1
- Inspection effectiveness goes from D1 to A1
- ar/t factor becomes 0.31
- DF changes from 145 to 7 (interpolated)
- Next internal inspection when DF reaches 85, this is when ar/t reaches 0.54, in the year 2027

API trajectory (inspection)



API trajectory (inspection)

ar/t	Number of Inspections and combined credit																
	0	1	2	3	4	1	2	3	4	1	2	3	4	1	2	3	4
	E	A				B				C				D			
0.05	4	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
0.10	14	1	1	1	1	1	1	1	1	1	1	1	1	3	2	1	1
0.15	32	1	1	1	1	1	1	1	1	2	1	1	1	8	4	2	1
0.20	56	1	1	1	1	2	1	1	1	6	2	1	1	18	10	6	3
0.25	87	1	1	1	1	4	1	1	1	11	4	2	1	32	19	11	7
0.30	125	1	1	1	1	9	3	1	1	21	9	4	2	53	33	21	14
0.35	170	1	1	1	1	16	5	2	1	36	16	8	4	80	54	36	24
0.40	222	2	3	1	1	29	10	4	2	57	29	14	7	141	81	57	40
0.45	281	3	6	1	1	47	19	8	3	86	47	26	14	202	124	86	63
0.50	347	4	10	2	1	73	33	15	7	124	73	43	20	273	217	173	138
0.55	420	5	15	3	1	109	55	28	14	173	109	69	44	346	285	234	192
0.60	500	6	21	5	1	158	88	49	28	234	158	107	73	430	365	309	262
0.65	587	7	27	7	1	222	136	83	51	309	222	160	115	527	459	401	349
0.70	681	8	34	9	2	305	202	134	89	401	305	232	177	635	569	510	456
0.75	782	9	42	11	2	409	294	211	152	510	409	328	263	757	695	638	586
0.80	890	10	51	13	3	538	416	322	249	638	538	453	382	893	839	789	741
0.85	1005	11	61	16	3	696	578	479	397	789	696	615	543	1044	1002	963	925
0.90	1126	12	72	19	4	888	787	697	617	963	888	819	756	1209	1186	1163	1140
0.95	1255	13	84	23	4	1118	1054	993	936	1163	1118	1075	1033	1390	1390	1390	1390
1.00	1390	13	97	27	4	1390	1390	1390	1390	1390	1390	1390	1390	1390	1390	1390	1390

Client case tank T44 (maintenance)

What if the client decides to do maintenance works on the tank and on the tankpit?

Present situation:

- The tankpit is too small to contain all the product in case of rupture
- The groundwater level is very high
- There is no leak prevention barrier on or under the tankfloor

Client case (maintenance)

- New and higher tank dikes are drawn up from concrete
- The tankpit is covered in concrete to keep groundwater out of the tankpit
- A RPB is installed (for instance coating or a double floor)

API 581 re-evaluation (maintenance)

- CoF bottom leaks becomes 1130 dollars
- CoF rupture for floor and shell becomes 1772499 dollars
- CoF shell leaks becomes 1987 dollars
- CoF financial for shell leaks remains the same
- CoF financial for shell rupture remains the same
- CoF financial for bottom leaks remains the same
- CoF financial for bottom rupture remains the same

Api 581 re-evaluation (maintenance)

- Weighted CoF shell goes from category E to category C
- Weighted CoF floor goes from category E to category C
- Acceptable damage factor floor becomes 250, the calculated damage factor was 145.
- The next inspection date is when ar/t reaches 0.53; this is in the year 2024.

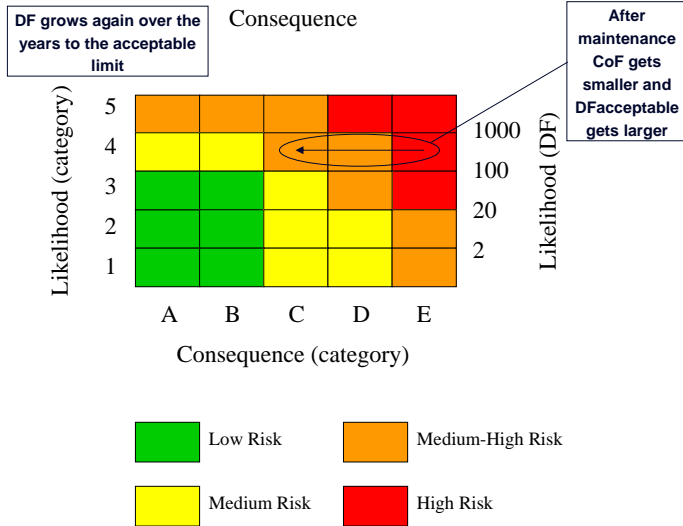
API trajectory (maintenance)

ar/t	Number of Inspections and combined credit																
	0	1	2	3	4	1	2	3	4	1	2	3	4	1	2	3	4
	E	A				B				C				D			
0.05	4	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
0.10	14	1	1	1	1	1	1	1	1	1	1	1	1	3	2	1	1
0.15	32	1	1	1	1	1	1	1	1	2	1	1	1	8	4	2	1
0.20	56	1	1	1	1	2	1	1	1	6	2	1	1	18	10	6	3
0.25	87	3	1	1	1	4	1	1	1	11	4	2	1	32	19	11	7
0.30	125	6	1	1	1	9	3	1	1	21	9	4	2	53	33	21	14
0.35	170	12	1	1	1	16	5	2	1	36	16	6	3	87	54	36	24
0.40	222	21	3	1	1	29	10	4	2	57	29	11	5	114	81	57	40
0.45	281	36	6	1	1	47	19	8	3	86	47	18	14	154	117	86	63
0.50	347	58	12	3	1	73	33	15	7	124	73	27	21	217	162	124	95
0.55	420	89	23	6	2	109	55	28	14	173	109	36	27	273	217	173	138
0.60	500	133	41	13	4	158	88	49	28	234	158	45	36	346	285	234	192
0.65	587	192	71	27	10	222	136	83	51	309	222	60	45	430	365	309	262
0.70	681	270	119	53	23	305	202	134	89	401	305	81	54	527	459	401	349
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0.90	1126	856	672	527	414	888	787	697	617	963	888	191	90	1044	1002	963	925
0.95	1255	1098	976	867	771	1118	1054	993	936	1163	1118	222	100	1209	1186	1163	1140
1.00	1390	1390	1390	1390	1390	1390	1390	1390	1390	1390	1390	255	100	1390	1390	1390	1390

Present situation tank T44

Damage factor grows to 250 over the years, this triggers the next inspection.

API trajectory (maintenance)



Conclusions tank T44

- First RBI study yielded high damage factor because of insecurities in inspection data
- After the first inspection cycle the damage factor is much smaller (learning effect!) and the inspection term much larger
- Apart from inspections, maintenance has a very positive effect on the integrity and the RBI results
- Tank shell didn't need inspection, so no unnecessary inspections



TankFarm Integrity Conclusions

- Turn-key service offered by RTD, also includes piping
- Cost-savings through onstream inspections
- Cost-savings through inspection interval extension
- RBI is a strong planning tool
- Easy to build inspection budgets
- Less unplanned outages
- Less unplanned maintenance and inspection
- Better integrity knowledge of the tanks



TankFarm Integrity Conclusions

- Inspections become more specific and tailor-made
- RBI also takes maintenance into account

Appendix 12

In-service equipment integrity

Jacco Rosendaal (RTD Group)



RTD
QUALITY SERVICES

In-service Integrity Inspection

J.F Rosendaal - March 2006 - NACE Italy

RTD
QUALITY SERVICES

Contents

- In-service Integrity Inspection
- Know the possibilities & limitations
- Major fields of application:
 - Application: Pre-shutdown
 - Application: RBI
- In-service NDT methods
- Conclusion

In-service Integrity Inspection



In-service Integrity Inspection

Means:

- Inspect whenever you want while in-service
- No unnecessary destruction of insulation or other
- Status on interior/exterior of equipment

In-service Integrity Inspection



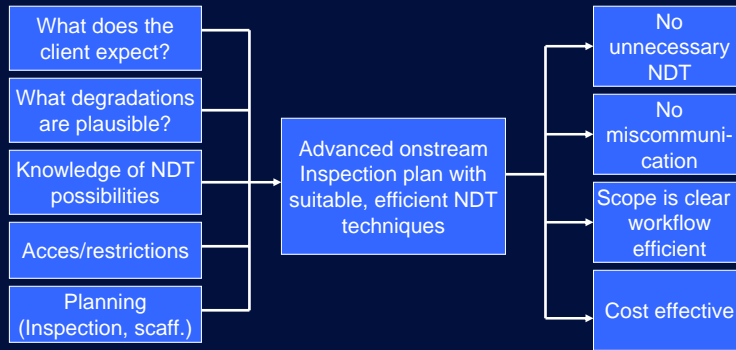
Possibilities & limitations Understanding the request

- There is no single NDT method to find all problems. Every method has charms & issues.
- What exactly is the problem? What does the client expect?
- What can our provider do & what does he needs to know / needs prepared?
- Discussion between provider & client!
- Pre-inspection, quickscan, etc...

In-service Integrity Inspection



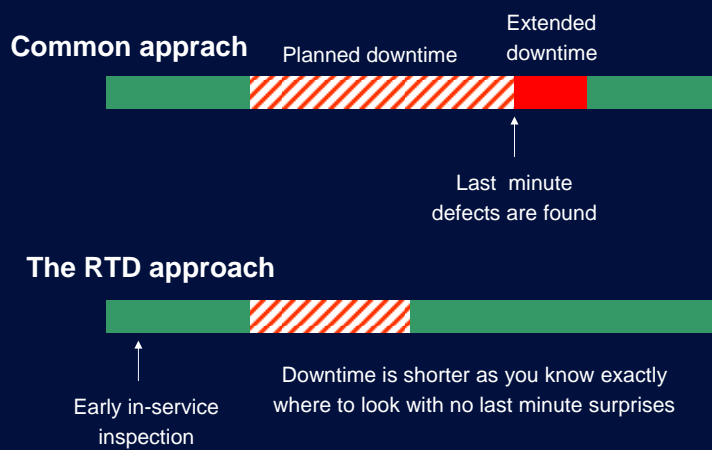
Possibilities & limitations Construct a solid plan



In-service Integrity Inspection



Application: Pre-shutdown Minimising Downtime



In-service Integrity Inspection



Application: Pre-shutdown Minimising Downtime

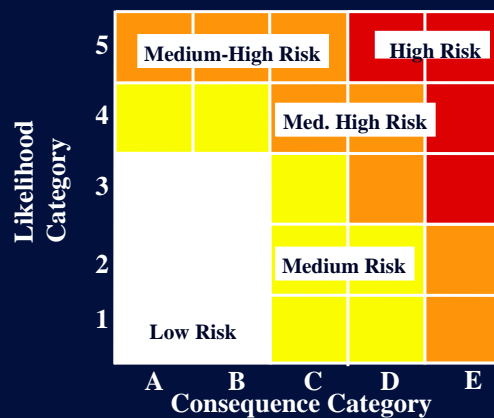
- Severe problem >> Know where to look in shutdown and perform effective maintenance
- Minor problem >> monitor and postpone maintenance
- Clean bill of health >> leave as it is

In-service Integrity Inspection



Application: Lowering RBI risk Evaluate riskfactor by facts

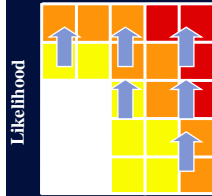
Level of NDT = Level of Risk



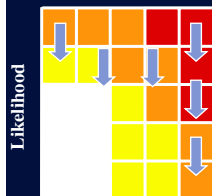
In-service Integrity Inspection



Application: Lowering RBI risk Adjust riskfactor to latest facts



Likelihood of failure will increase over time because of time-dependent material degradation

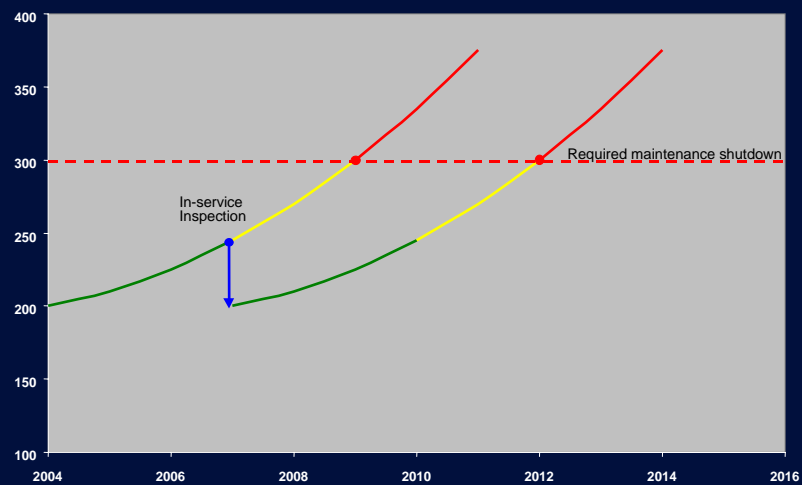


The likelihood of failure can be reduced by improved inspection planning and good inspection methods

In-service Integrity Inspection



Application: Lowering RBI risk Postpone required shutdown



In-service Integrity Inspection

In-service NDT methods: Screening and Measuring

■ Screening:

Inspect large sections of equipment relatively fast with low detail

- Method only indicates there is something out of the ordinary, nothing more

■ Measuring:

Inspect small sections of equipment relatively slow with high detail

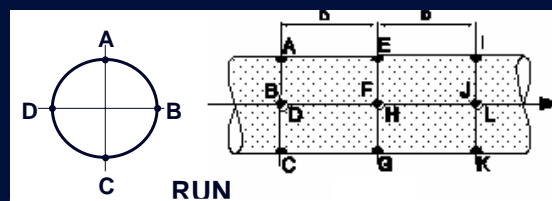
- Method gives accurate figures and numbers

In-service Integrity Inspection

In-service Screening: Equivalent of UT spot check

What is UT spot check?

- Wallthickness measurement at one single point.
- Mapping according to grid
- Only global degradation; detection of local corrosion is sheer luck



In-service Integrity Inspection



In-service Screening: Equivalent of UT spot check

What is RTD Incotest?

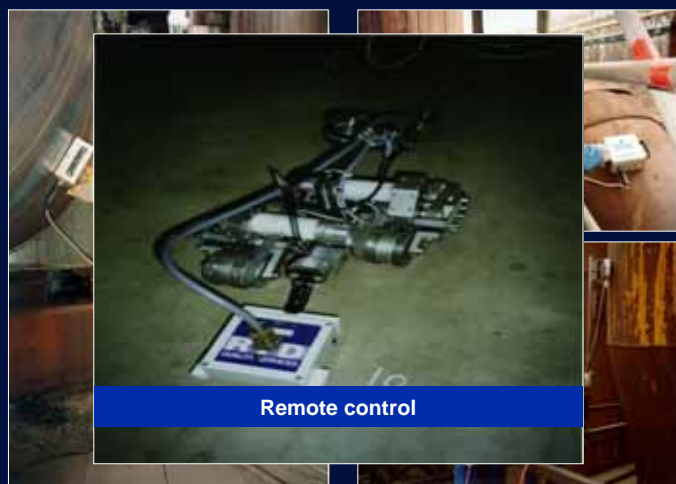
- Wallthickness screening in one single area.
- Mapping according to grid
- Only global degradation; detection of local corrosion is sheer luck



In-service Integrity Inspection



In-service Screening: Equivalent of UT spot check



In-service Integrity Inspection



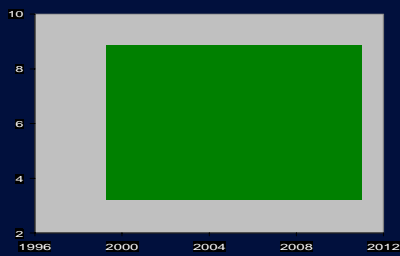
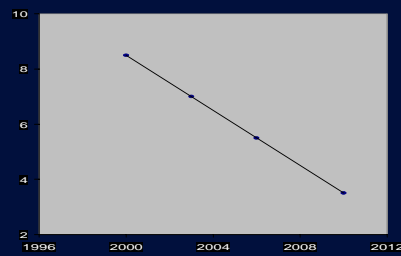
In-service Screening: Equivalent of UT spot check

UT spotcheck

- Requires removal of insulation
- Measurements once during shutdown period
- Gives exact wallthickness

RTD-Incotest

- In-service through insulation
- Measuring whenever customer sees fit
- Gives average wallthickness



In-service Integrity Inspection



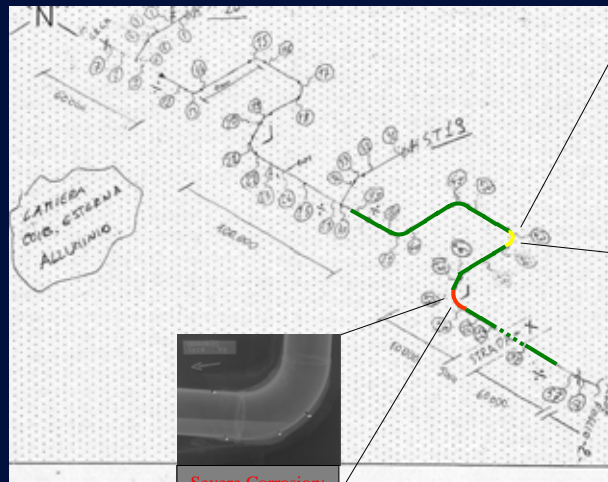
In-service Screening: Equivalent of UT spot check



In-service Integrity Inspection



In-service Screening: Screening for suspicious locations



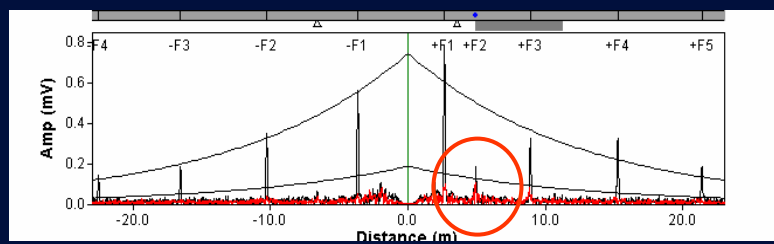
Severe Corrosion!

False Call

In-service Integrity Inspection



In-service Screening: Screening for suspicious locations



In-service Integrity Inspection



In-service Screening: Screening for suspicious locations



Pipebridge



Jetty lines / piperacks



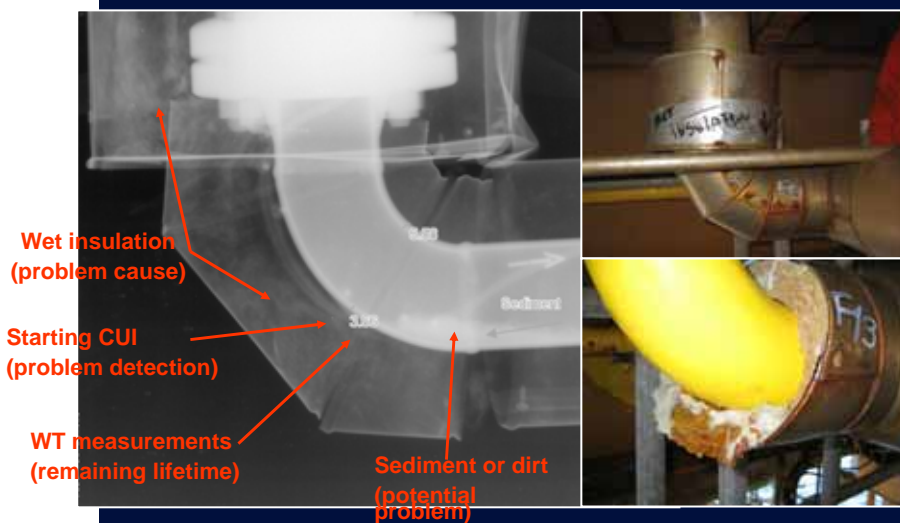
Roadcrossing with sleeve



In-service Integrity Inspection



In-service measuring: Detailed (Follow-up) Inspection



In-service Integrity Inspection

In-service measuring: Detailed (Follow-up) Inspection



In-service Integrity Inspection

In Conclusion: In-service Integrity Inspection

Means:

- Inspect whenever you want while in-service
- No unnecessary destruction of insulation or other
- Status on interior/exterior of equipment

- Decrease downtime - know where to look!
- No last minute surprises

- No problems? >> lower RBI risk >> extend operation period

In-service Integrity Inspection



Thank you for your attention

In-service Integrity Inspection

Appendix 13

Artificial Neural Network for process control and monitoring

Giovanni Zangari (Process control consulting)

ARTIFICIAL NEURAL NETWORK FOR PROCESS CONTROL AND MONITORING

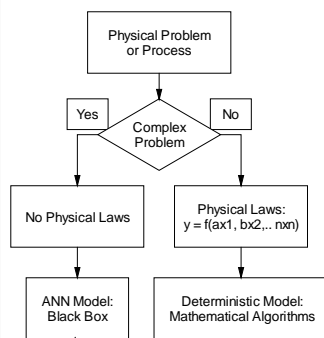
Aim of the presentation

Introduction to the application of Artificial Neural Networks (ANNs) in industrial process control and monitoring.

- General consideration about ANNs methodology.
- Examples taken from steel making industry;

The approach to the problem is highlighted to demonstrate the exportability of such methodology in other industrial fields, also in petrochemical and refinery one Then, an overview of practical applications will be presented, highlighting the approach.

ARTIFICIAL NEURAL NETWORK FOR PROCESS CONTROL AND MONITORING



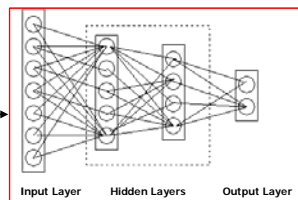
A physical phenomenon is described by a set of variables correlated by physical laws:

$$y = f(ax_1, bx_2, \dots, nx_n)$$

ANN represents an extreme and flexible attempt to codify the relationship between a set of variables (real, discrete and boolean).

ANNs application field:

- Image and speech;
- Banking and business
- Medical instrumentation
- Industrial process
- Social and biology



ARTIFICIAL NEURAL NETWORK FOR PROCESS CONTROL AND MONITORING

Condition for a suitable application of an ANN model:

- hard definition and calibration of a deterministic mathematical model
- unsatisfactory performance of other techniques (i.e. Linear and Not-Linear multiple regression)
- not exportability of similar model

Requirements for the application of an ANN model:

- Availability of a set of process observation enough to cover the range of variability of the variables

3

ARTIFICIAL NEURAL NETWORK - ANN Required Procedures in the Development of an Artificial Neural Network Model

The list below presents the main activities carried out during the development of an ANN in a suitable order. Some of them are iteratively repeated to achieve the best performance.

- Process variables definition and collection (Input / Output)
- Normalization
- Process variables selection (PCA)
- Process data filtering (out-layer selection)
- Sample extraction to obtain a more uniform distribution
- Training, testing and model definition

4

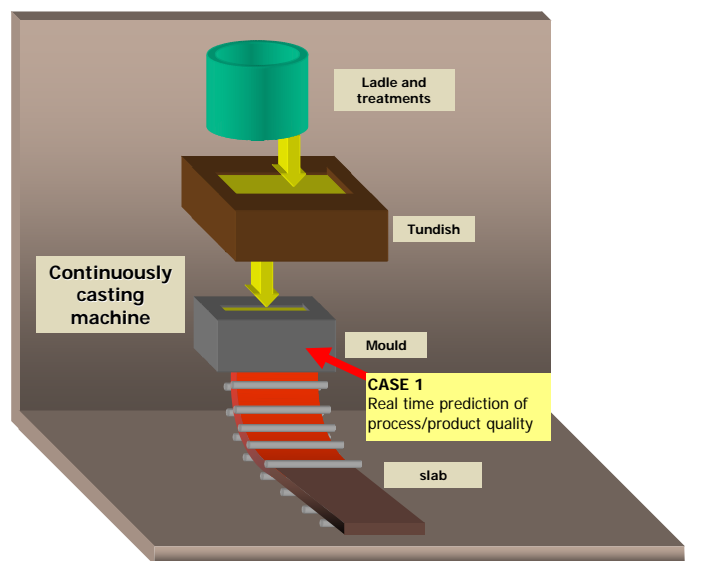
ARTIFICIAL NEURAL NETWORK - ANN INDUSTRIAL APPLICATIONS

- **CASTING STEEL** (case 1: **SOLIDIFICATION**)
- **ROLLING MILL** (case 2, 3, 4: **THERMAL EVOLUTION**)
- **IMAGE CLASSIFICATION** (case 5: **STRIP DEFECT**)
- **TIME SERIES** (case 6: **ELECTRIC POWER DEMAND**)
- **VIRTUAL INSTRUMENT** (case 7: **ORGANIC COMPOUNDS CONTENT**; case 8: **MECHANICAL PROPERTIES OF STEEL**)
- **CORROSION CLASSIFICATION** (case 9: **IDENTIFICATION OF LOCALIZED CORROSION BY ELECTROCHEMICAL NOISE - work in progress**)

5

CASTING STEEL (MELTING AND SOLIDIFICATION)

SCHEME OF ELECTRIC ARC FURNACE AND CONTINUOUSLY CASTING MACHINE



6

CASTING STEEL (MELTING AND SOLIDIFICATION)

CASE 1: REAL TIME PREDICTION OF PROCESS (PRODUCT) QUALITY

The continuous casting process is described by time dependent variables. A reference data set is made of sample of only correct processes. If an unknown sample is similar to one (or more) from the reference data set, the probability that it regards a correct process increases (otherwise the process should be not correct, not steady or not frequent).

An Associative Memory based on a ANN is trained to codify correct process sample (transient or unspecified process condition are neglected). The network is trained to reproduce in output the same input pattern.

The RMS error is calculated from the difference between the input and output. This identifies a process quality index.

The higher the index (the lower the RMS error), the higher the probability that the process is similar to a correct one.

Input Process Variables

Steel level in the mould:

- Standard deviation of mould steel level oscillation
- mould steel level measure
- time trend of mould steel level

Mould wall temperature

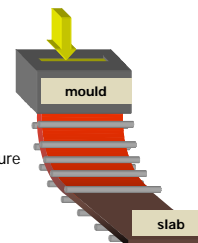
- Measured temperature
- Standard deviation of temperature

Cooling water:

- Standard deviation of flow rate
- Standard deviation of delta temperature

Casting speed:

- Standard deviation



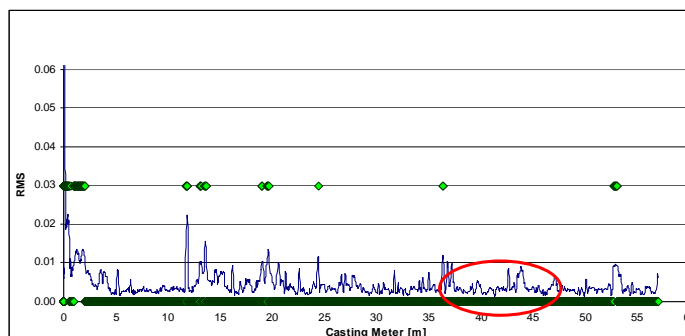
CASTING STEEL (MELTING AND SOLIDIFICATION)

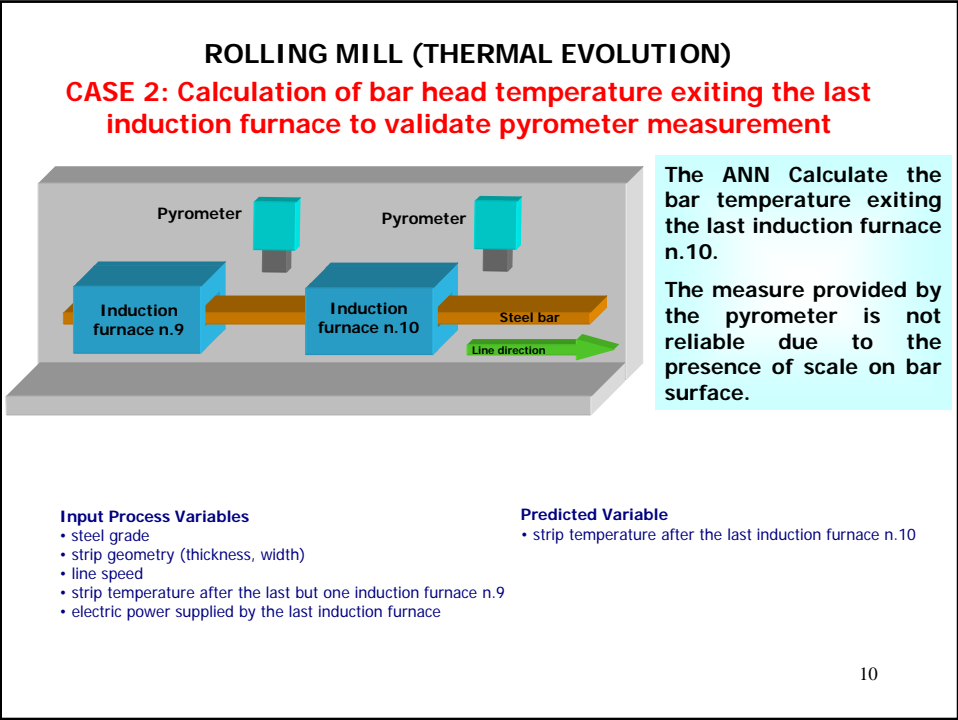
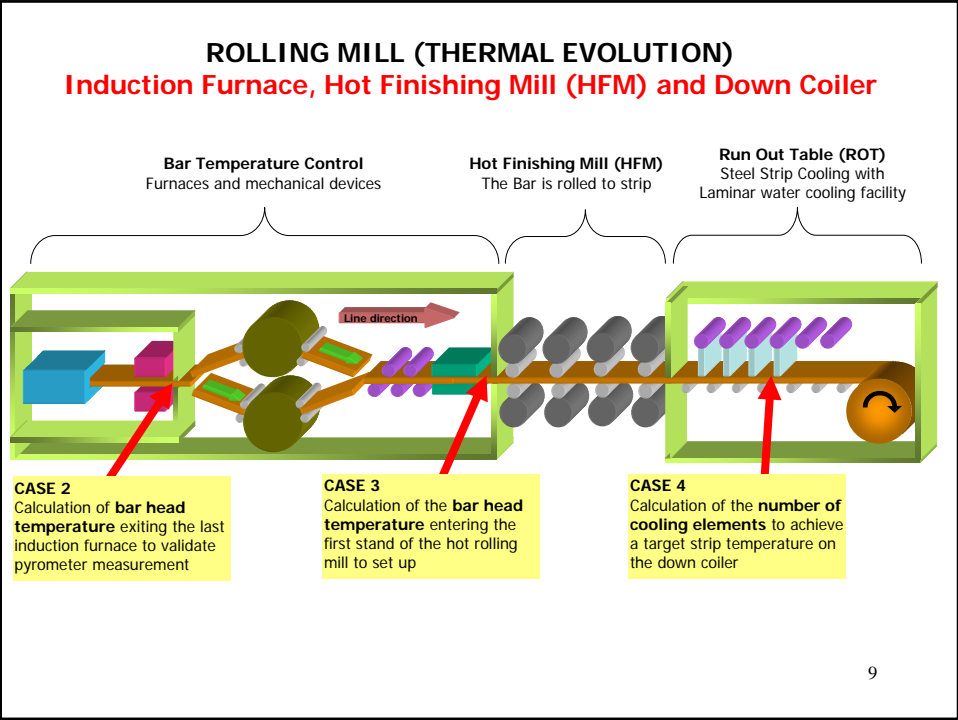
CASE 1: REAL TIME PREDICTION OF PROCESS/PRODUCT QUALITY

The red circles reasonably identify the time when a product with defect occurred that can be associated to not steady state process condition.

The on line version produce the signal below; in most of cases these are related to low quality products

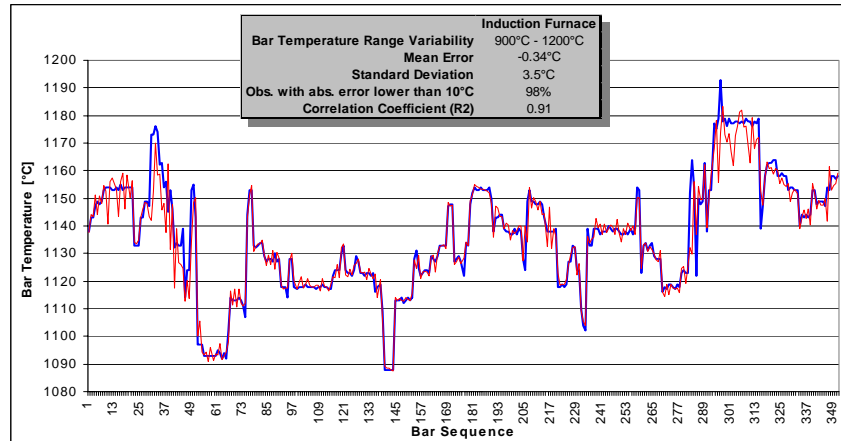
The green squares identifies the ANN model response with an RMS error above a threshold.



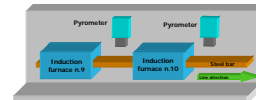


ROLLING MILL (THERMAL EVOLUTION)

CASE 2: Calculation of bar head temperature exiting the last induction furnace to validate pyrometer measurement

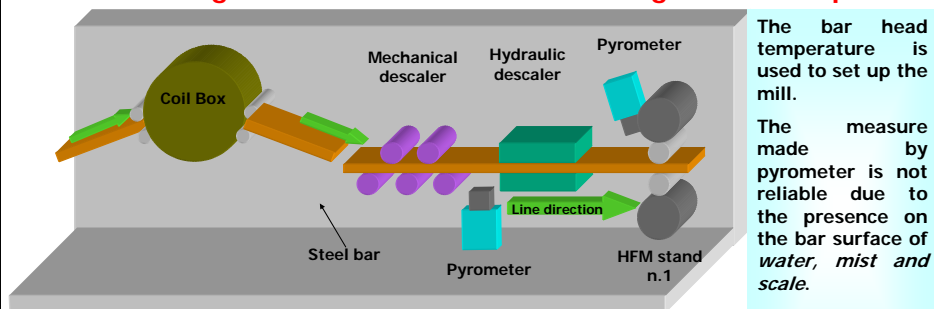


Temperature of the bar out of the Induction Furnace n.10 measured (blue line) vs. predicted (red line) with the Neural Model



ROLLING MILL (THERMAL EVOLUTION)

CASE 3: Calculation of the bar head temperature entering the first stand of the Hot Rolling Mill to set up



The bar head temperature is used to set up the mill.

The measure made by pyrometer is not reliable due to the presence on the bar surface of *water, mist and scale*.

Moreover the measure is not available in time for setting up the mill.

Input Process Variables

- steel grade
- strip geometry (thickness, width)
- rolling speed
- strip temperature after the last induction furnace
- bar time in the coil box
- inner temperature of the coil box
- hydraulic descaling pressure
- mechanical descaling on/off
- Burners on/off

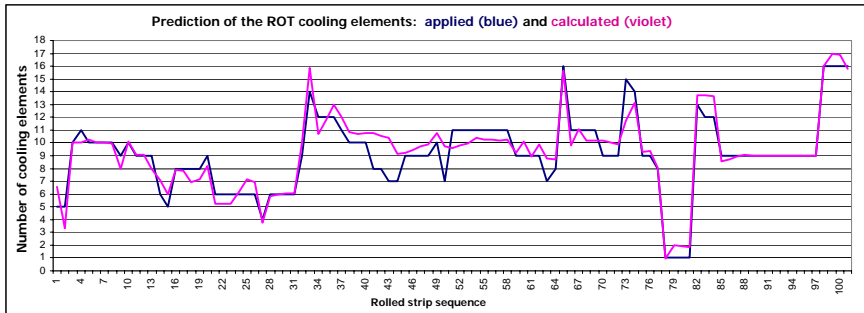
Predicted Variable

- bar temperature entering the hot rolling mill

An ANN model acts as a *virtual pyrometer* providing a more reliable and in-time estimation of the bar head temperature .

ROLLING MILL (THERMAL EVOLUTION)

CASE 4: Calculation of the number of cooling elements on the ROT to achieve a target strip temperature on the down coiler.



The on-line version of the ANN calculates the number of water cooling elements with an error lower than 2 elements in the 94% of process observations.

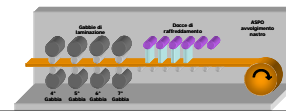
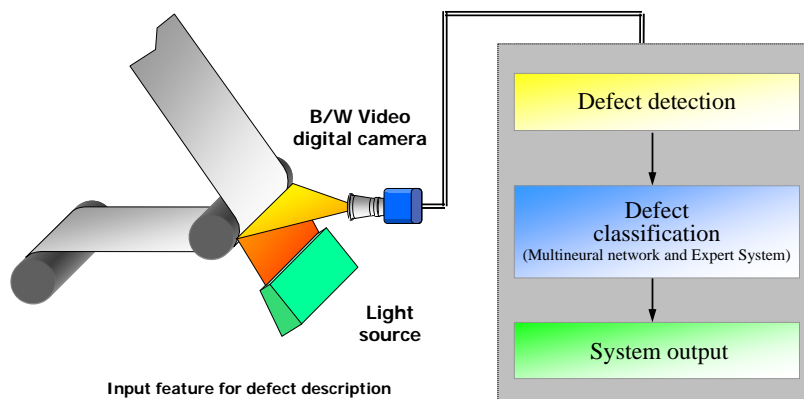


IMAGE CLASSIFICATION

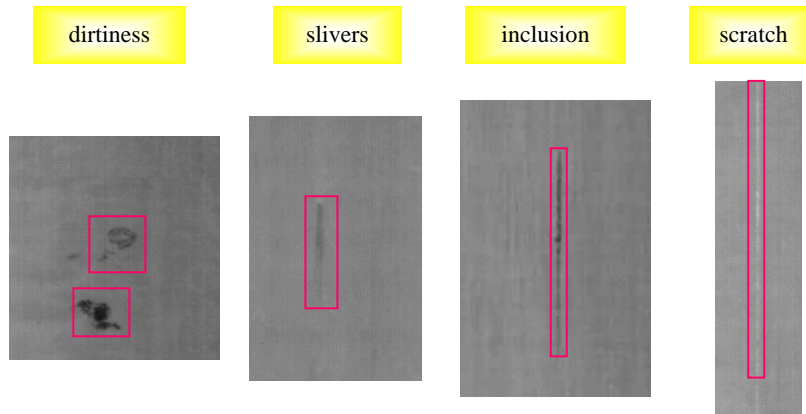
CASE 5: CLASSIFICATION OF STRIP SURFACE DEFECT PROVIDED BY AN INSPECTION SYSTEM



- Input feature for defect description**
1. Steel grade
 2. defect geometrical dimension
 3. picture contrast in grey scale
 4. two-dimensional elongation ratio
 5. cyclic phenomenon and frequency
 6. Strip position

IMAGE CLASSIFICATION
CASE 5: CLASSIFICATION OF STRIP SURFACE DEFECT PROVIDED BY AN INSPECTION SYSTEM

Example of strip surface defects

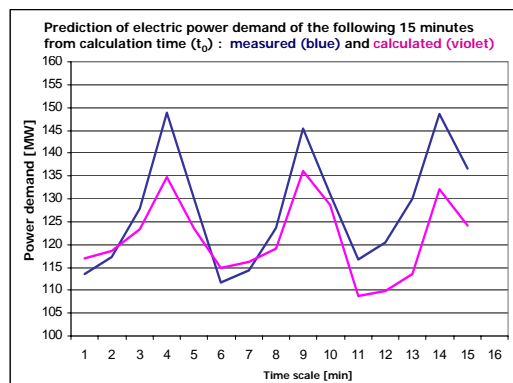


TIME SERIES
CASE 6: PREDICTION OF ELECTRIC POWER DEMAND OF A STEEL WORKS

An ANN model performs a time series calculation predicting the electric power demand of a steel work plant. The input pattern is the power demand of the last 60 minutes and calculate the power demand of the following minutes.

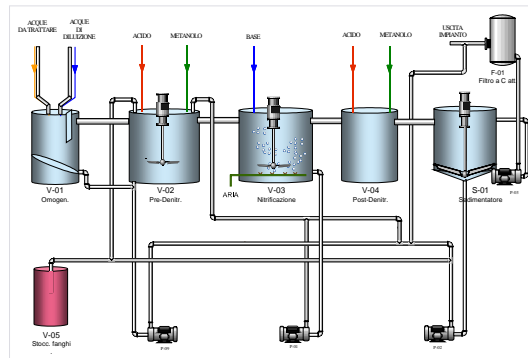
The on line version of the ANN model calculates the power demand of the further 5 minutes with the following average error.

[min]	[MW] error
1	5.9
2	7.5
3	7.8
4	7.9
5	8.2



VIRTUAL INSTRUMENT

CASE 7: PREDICTION OF TOC/BOD AND CHROMIUM CONTENT FROM A WASTE WATER TREATING PLANT



The TOC/BOD or chromium content of industrial water after treatment can not be a fast measure to perform.

As a consequence a waste water treating process is not easy to control.

An ANN model acts as a virtual instrument providing a real time estimation of the organic compound content.

Input feature for defect description

1. Inlet and outlet water temperature
2. Inlet water flow rate
3. Inlet water TOC/BOD content
4. Operative condition
5. Chemical additive in treatment tanks

Predicted Variable

- Outlet TOC/BOD or chromium content

VIRTUAL INSTRUMENT & VALIDATION

CASE 8: PREDICTION OF THE MECHANICAL PROPERTIES OF HOT ROLLED STEEL STRIP (work in progress)

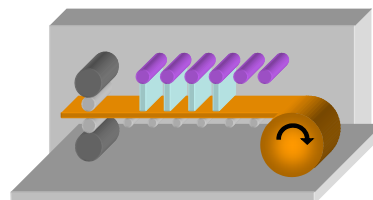
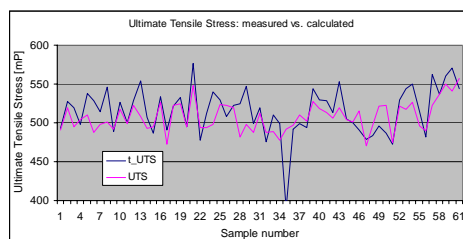
The mechanical properties (Yield Stress, Ultimate Tensile Stress, Elongation and Hardness) of hot rolled steel strip are a function of the rolling and cooling operative procedures.

Input variables:

- Steel chemical composition
- Strip thickness
- Strip speed out of the last rolling stand
- Strip thermal evolution during coiling

Output variables:

- Yield Stress
- Ultimate Tensile Stress
- Elongation
- Vickers Hardness



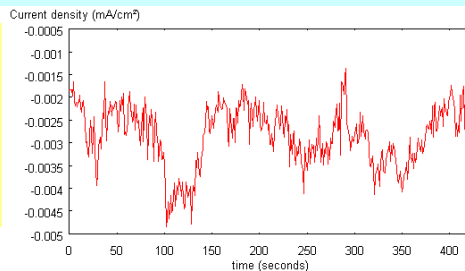
CORROSION CLASSIFICATION

CASE 9: IDENTIFICATION OF LOCALIZED CORROSION USING ELECTROCHEMICAL NOISE (work in progress)

The presence of Localized corrosion is identified by analysing the complete electrical signal (current and potential) collected in a time window, its elementary statistical indexes and the parameters provided by FFT performing a multi-variate analysis with a sequence of two Artificial Neural Networks.

The records processed by the ANNs are made of the following information:

- Direct Current signal
- Direct Potential signal
- Slope of interpolating 1st order curve
- Standard deviation
- Skewness
- Kurtosis
- FFT-parameters



The first ANN is an Associative Memory. It classifies uniform and localized corrosion sample: for this reason the ANN is trained only with uniform corrosion samples. The RMS error made by the ANN, processing an unknown sample, identifies the probability to be a localized corrosion sample (applying similarity criteria).

Once a case of not uniform corrosion is identified, a second ANN classifies the type of localized corrosion (pitting, crevice and others).

DESCRIPTION OF THE MAIN ARTIFICIAL NEURAL NETWORK FAMILIES

ANN FAMILIES

- Multi-layer Perceptron using Back Propagation training algorithm (MLP-BP)
- Radial Basis Function (RBF)
- Learning Vector Quantifier (LVQ)
- Constraint Satisfaction Associative Memory (ANN-AM)

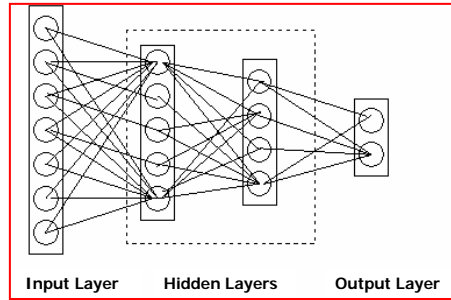
**DESCRIPTION OF THE MAIN
ARTIFICIAL NEURAL NETWORK FAMILIES**

**Multi-layer Perceptron Network using Back
Propagation training algorithm (MLP-BP)**

The MLP-BP neural network is employed in both classification problems (producing Boolean and integer output number) and in the prediction of real variables (with real type output number).

The ANN model is used in **recall** presenting the input pattern on the nodes of the input layer and projecting their activation signal to the next layers, up to the output layer.

The model **training** procedure concerns the back-ward correction of the connections (weights) between nodes of different layers starting from the output layer to the input one knowing the error between the network output and the target. The knowledge of the physical phenomenon is then codified on the weights.



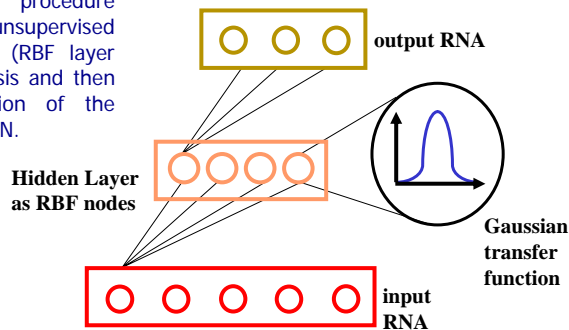
1. Recall ➔
2. Error calculation ↔
3. Weight correction ↻

**DESCRIPTION OF THE MAIN
ARTIFICIAL NEURAL NETWORK FAMILIES**

Radial Basis Function (RBF)

The RBF neural network is similar to the BP-ANN regarding the application fields (Boolean discrete and real variables) and the inner structure also including an extra layer containing RBF nodes corresponding to the centroids of a Cluster Analysis.

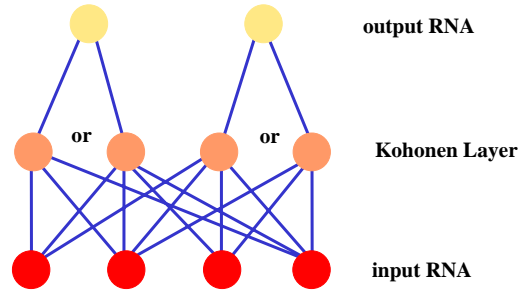
The model **training** procedure concerns first the unsupervised calculation of centroids (RBF layer nodes) by Cluster Analysis and then the back-ward correction of the weights as for the BP-ANN.



DESCRIPTION OF THE MAIN ARTIFICIAL NEURAL NETWORK FAMILIES

Learning Vector Quantization (LVQ)

The LVQ network is suitable only for classification problem. It is similar to RBF_ANN because of the presence of the Kohonen layer containing nodes calculated with Cluster Analysis. The nodes of the Kohonen layer identify an attractive pole or prototypes of a class.

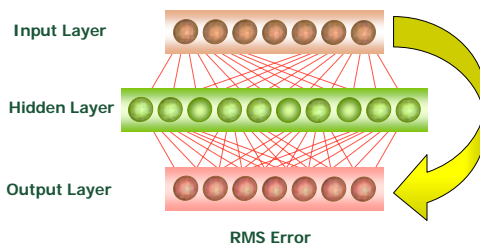


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DESCRIPTION OF THE MAIN ARTIFICIAL NEURAL NETWORK FAMILIES

Constraint Satisfaction Associative Memory (ANN-AM)

An **Associative Memory** codifies on the weights of a BP ANN family the association between the sample of a data set. It is trained to reproduce in output the same pattern presented in input. The recall phase can be performed in two different strategies.



Case 1: evaluation of the *affinity* of an unknown observation with the data set used in the ANN-AM training: the RMS error can be considered the affinity index: the higher the error, the lower the affinity.

Case 2: in case an input pattern has affinity with the training set and the value of a variable is lost, this can be evaluated with a recursive recall of the same pattern.

MAIN ITEMS INVOLVED IN THE TRAINING AND TEST OF NEURAL MODELS

Phase 2: Selection of the neural structure

Phase 3: Selection of the train and test data set

Phase 4: Scheduling of the training procedure

Phase 5: Plot of the RMS error and classification rate during training

Any ANN developing Tool provides the possibility to perform the activities listed in the figures on the left.

Once the training phase is over, the ANN model is put into ANSI-C source code, ready to be integrated with other existing codes.