



SUMMARY

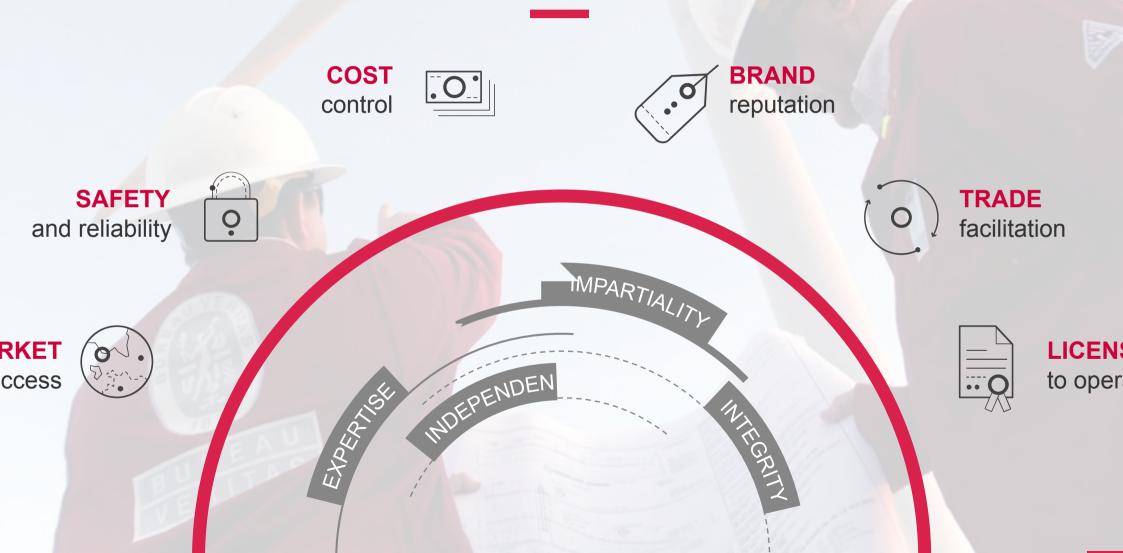
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BUREAU VERITAS
INTRODUCTION

DIGITAL STRATEGY & AI EFFORTS

03
BOILER PRESSURE PARTS
RESULTS



A GLOBAL LEADER OF THE TEST, INSPECTION AND CERTIFICATION MARKET



BUREAU VERITAS IN KEY FIGURES



4.8 billion

Revenue in 2018



More than

75,000

employees



400,000

clients



More than

1,500

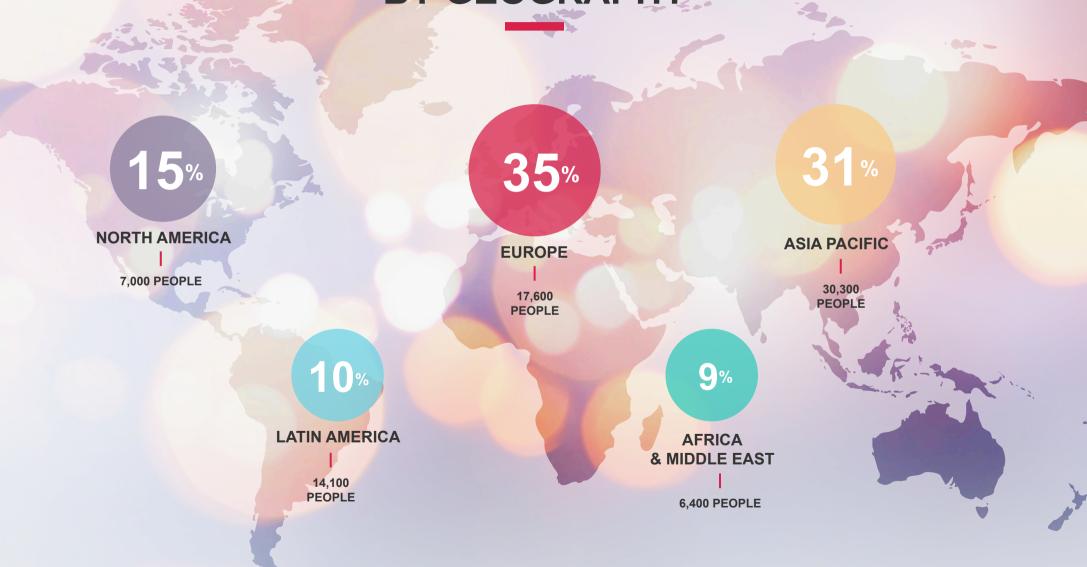
offices & laboratories in 140 countries



3,500

accreditations & agreements

REVENUE BREAKDOWN BY GEOGRAPHY









MARINE OFFSHORE

7%*

ety at sea through fshore platform in services. Provide opertise to assess e risks and improve e

INDUSTRY

22%*

Ensure the safety, security, reliability and integrity of industrial assets throughout their life cycle, and assess their compliance with national, international and voluntary QHSE standards. Control quality and provide supply chain optimization assistance in the automotive industry

* Percentage of 2018 revenue

BUILDINGS & INFRASTRUCTURE

27%*

Provide assurance that property assets and infrastructure in service or under construction are safe, energy efficient and comply with applicable regulations. Ensure business continuity and environmental protection by assessing the safety and efficiency of in-service installations and by analyzing air and water quality

CERTIFICATION

8%*

Certify that quality, health, safety and environmental management systems comply with applicable international, national or industry standards, or standards specific to large companies, in order to improve risk management and performance

AGRI-FOOD & COMMODITIES

22%*

Improve transparency, inspect the composition, quality and quantity of commodities throughout the value chain, from extraction to sale, and farm to fork. Facilitate international trade and protect citizens from poor quality products by verifying import conformity

CONSUMI PRODUC^{*}

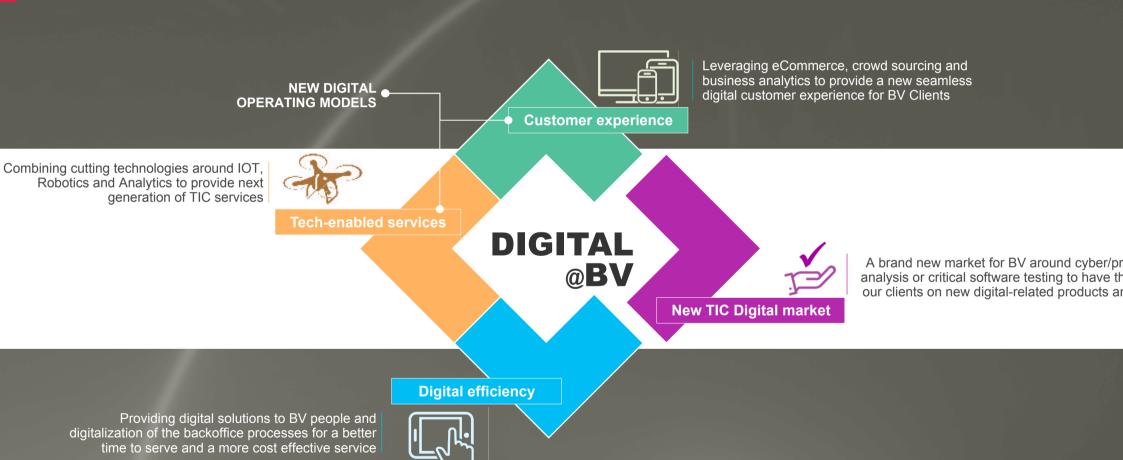
14%*

Test and verify consiproduct conformity, of safety and performa improve supply chain



DIGITAL STRATEGY & AI EFFORTS

JREAU VERITAS DIGITAL STRATEGY DIGITAL FRAMEWORK ON FOUR AXIS

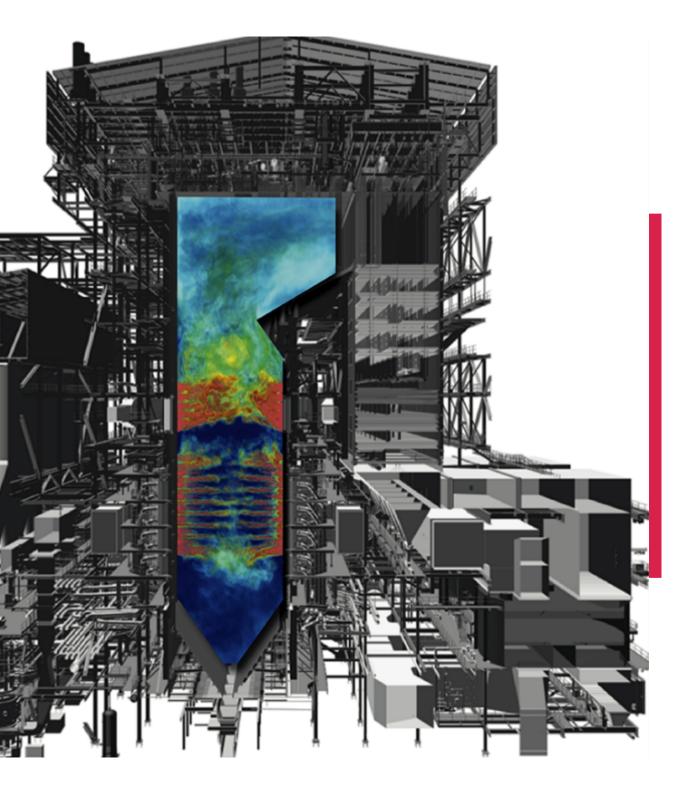


GITAL EFFICIENCY FOR LABORATORIES AB AUTOMATION

BV AUGMENTED LABS VIDEO



This will provide both our employees and our customers



COAL BOILER PRESSURE PARTS

Experimentation, results & methodology

INTEXT: PLANT CARE SOLUTION

er Pressure Parts includes all thin walled components ed to the boiler pressure envelope (e.g. superheaters, eaters, economizers and evaporators).

cal damage mechanisms: fretting, tube misalignment, ning from erosion, corrosion and visual inspection results.



essure parts failure have a big impact on plant availability and it is not always possible to do maintenance predefined scheduled times.

Pressure Parts module manages the inspection and repair process and provides the following features:

Data is QA checked with expected values & recaptured when needed to ensure quality

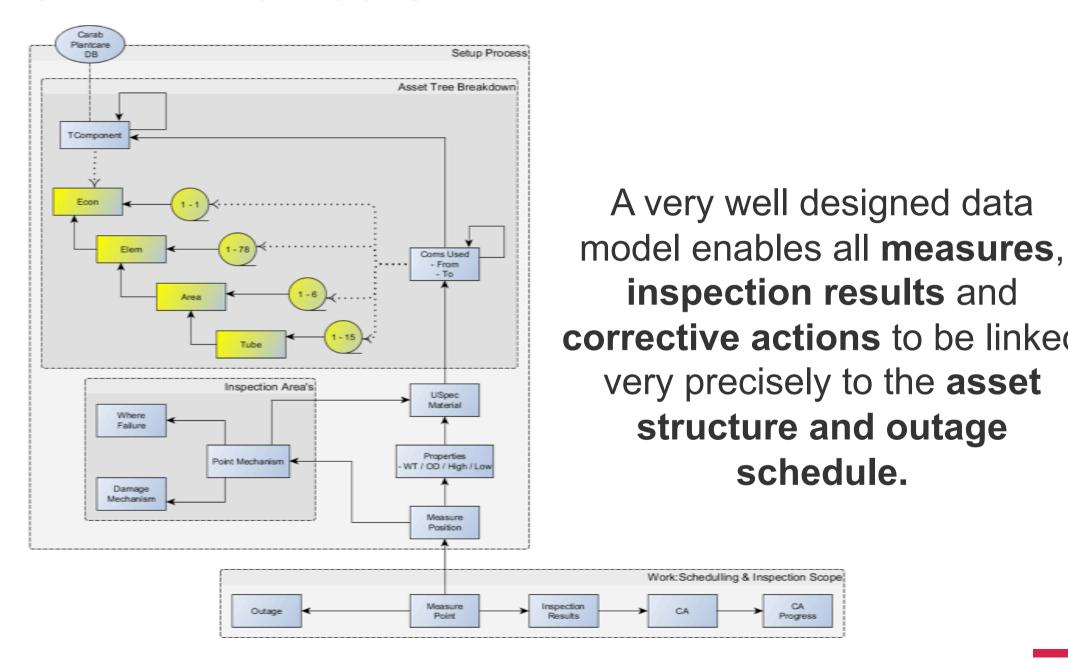
Wear Rates are calculated and the data is trended to asset the risk of cut lines

Budget scope is available before start of outage

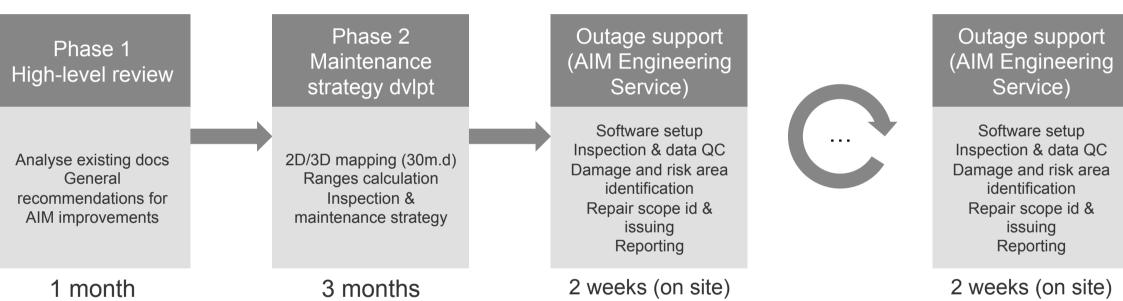
The flow of data is bi-directional and the original templates can be generated for final QA procedures.

The Boiler Pressure Parts module is used at 15 coal power stations.

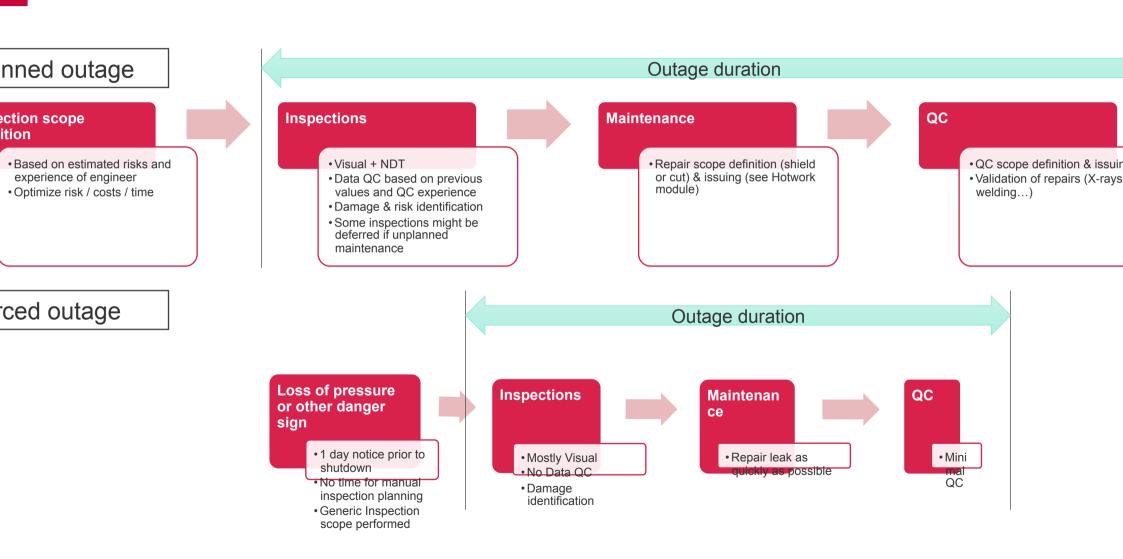
ANTCARE DATA STRUCTURE



ERVIEW OF THE CURRENT PLANTCARE AIM APPROACH



JTAGE SUPPORT PROCESS

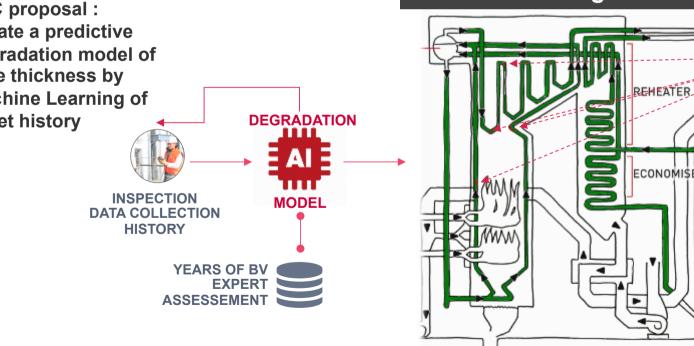


REDICTIVE RBI / AIM OAL BOILERS PRESSURE PARTS POC

Current situation

Despite a solid engineeringbased AIM approach, **some leaks still occur** Due to limits of current prediction methods, less than 20% of risk areas are actually inspected during an outage

Quality check of NDT measures is a lengthy and uncertain process



PREDICT VALUE FOR EACH
MEASURE POINT AND IDENTIFY
LEVEL OF RISK FOR EACH AREA

DETECT
WRONG
READINGS

IMPROVE
MODEL WITH
FINDINGS

C BUSINESS OBJECTIVES & KPIS

ectives

ouild a generic predictive model on the data for Site A Power station for the 6 Boiler units installed on the site and for 2 or components per unit viz. Economizer and Reheater.

e model should predict the tube thickness at any particular point in time. The model should also be able to identify risky as and also predict potential leaks.

uce effort and risk for asset outage & inspection planning		Identify >70% of risky areas in validation set
	1 .	
ess thickness prediction performance		Determination coefficient (R²) >90%
	1	
ess capability of model to support inspection data quality check		>90% measures should be in a >90% confidence interval from prediction
	-	
ance asset reliability with improved leak prediction and anticipated repair		At least one additional leak avoided as result of predictive model
	_	

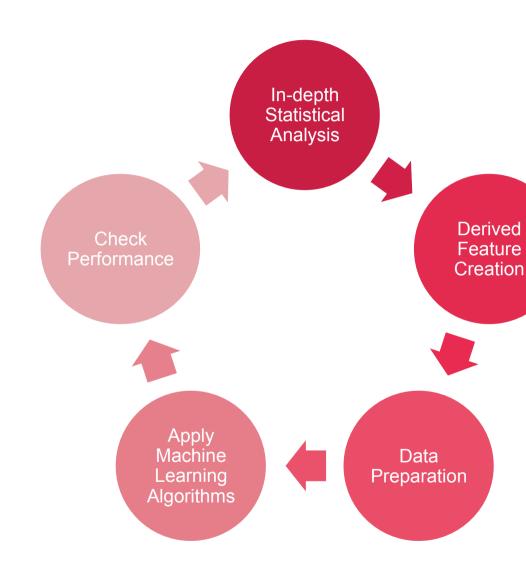
REDICTIVE MODEL DEVELOPMENT CYCLE

Data are split in separate learning & validation datasets so that predictive (future) performance of the model can be assessed

Learning Validation

8-9 years

1.5 years



LIDATION RESULTS REHEATER – FOR SITE A

Iteration 1 : better thickness prediction

KPI1:

	Predicted(1)	Predicted(0)
Actual(1)	444	181
Actual(0)	590	40378

71% Risks predicted

KPI 2: R2=0,74

KPI 3: 92% actual values lie in 90CL

KPI 4 : Predicted 1 out of 2 leakages

Iteration 2: better risk prediction

KPI 1:

	Predicted(1)	Predicted(0)	
Actual(1)	461	164	
Actual(0)	1047	39921	

74% Risks predicted

KPI 2 : **R2=0,67**

KPI 3:86% actual values lie in 85%CL

KPI 4 : Predicted 2 out of 2 leakages

Risk areas and leak prediction match success criteria.

Counter-intuitively, optimizing for risks detection decreases thickness prediction performance and vice-versa.

LIDATION RESULTS ECONOMIZER – FOR SITE A

Iteration 1 : better risk prediction

KPI 1:

	Predicted(1)	Predicted(0)
Actual(1)	283	83
Actual(0)	1088	45474

77% Risks predicted

KPI 2 : R2 = 31%

KPI 3:85% values lie in 80%CL

Iteration 2: better thickness prediction

KPI 1:

	Predicted(1)	Predicted(0)
Actual(1)	242	124
Actual(0)	720	45842

66% Risks predicted

KPI 2 : R2 = 37%

KPI 3:86% values lie in 80%CL

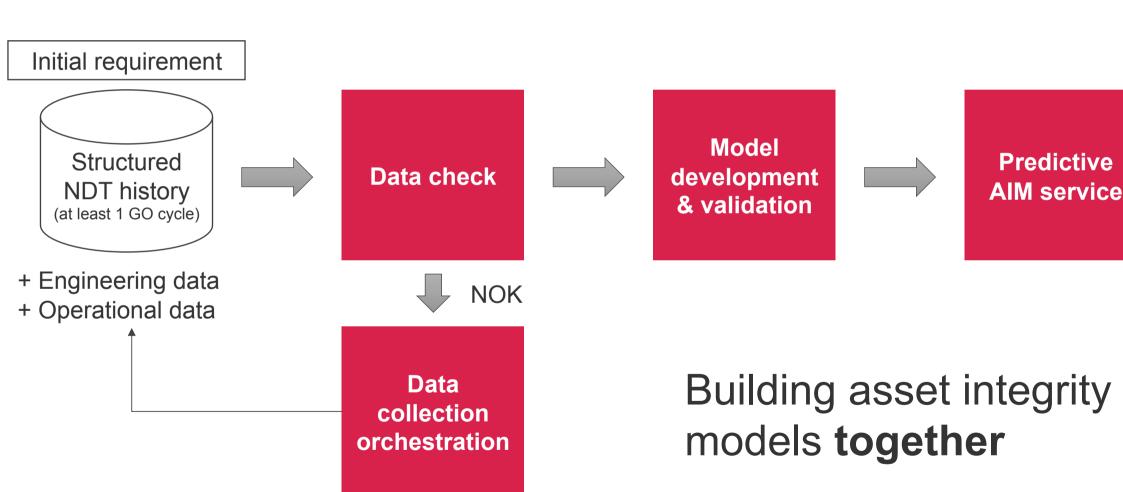
No leaks were present in the validation dataset

Due to less data than Reheater (and possibly worse data quality), thickness prediction performance is bad, but risk area prediction still works

JSINESS BENEFITS FOR CUSTOMER

	Current	New with model	Gain	
ne to plan a normal outage	5-10 days	1-2 days	4-8 m.days / outage (internal)	
al-time NDT ta Quality check	Limited, based on engineer experience	Systematic and automated	Better prediction (see below)	
k areas inspected in an age	<20%	>70%	Leak avoidance → reduced maintenance	
pacity to define a specific c-based plan for opportunity xt day) outage	NO	YES	costs + forced outage avoidance → 421 k€ / leak avoided (est. medium size unit + possible reduction of number of inspections)	
ak prediction	Yes, but with very significant human effort, and big probability of miss	Yes, result of the computation, with a good chance of finding		

ENERIC METHODOLOGY FOR OTHER ASSETS





THANK YOU

Michel-Ange Camhi

michel-ange.camhi@bureauveritas.com

