

Data quality improvement





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DIGITAL SOLUTIONS

Data Quality Improvement

Foundations of data quality improvement in data collection architectures

Simen Sandelien 18 September 2019

Speaker



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Topics

- DNV GL
- Digital Class and data historian infrastructures
- Data quality perspectives
- Sensor system data quality foundations
- Recommendations
- Questions

DNV GL - global quality assurance and risk management company

OUR PURPOSE

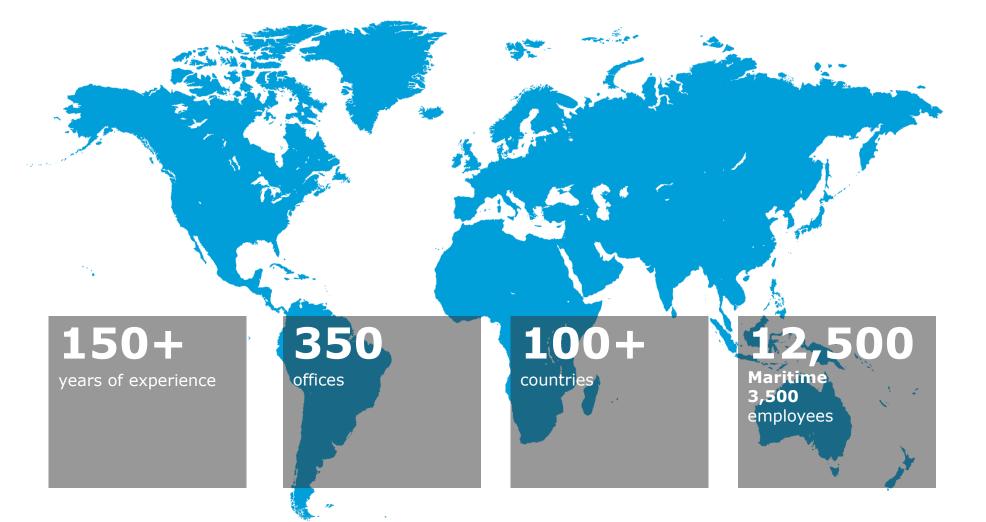
TO SAFEGUARD LIFE, PROPERTY AND THE ENVIRONMENT

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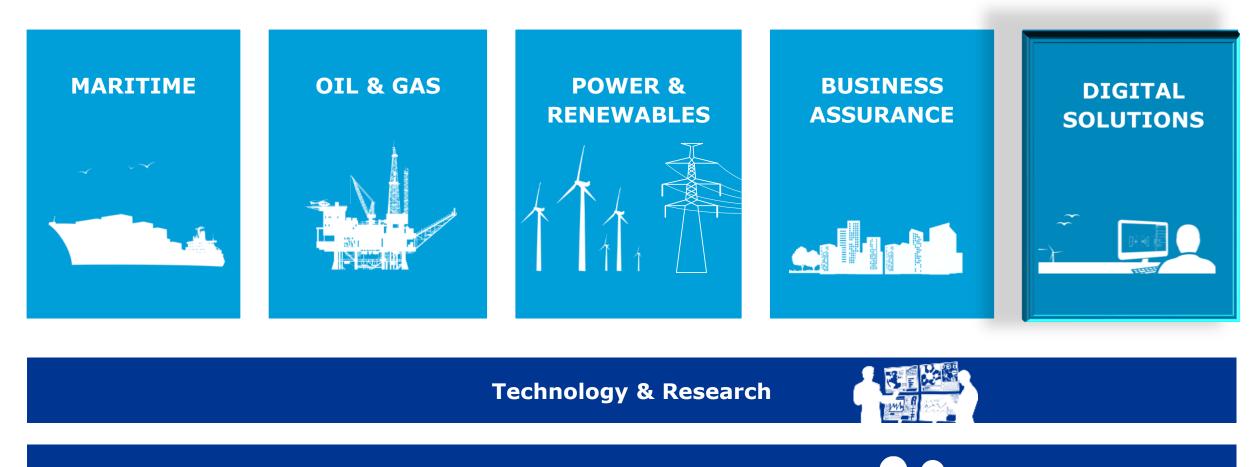
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Our vision: global impact for a safe and sustainable future



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Digital solutions for managing risk, improving safety and performance

We support business-critical activities across industries, including maritime, oil and gas, energy and healthcare

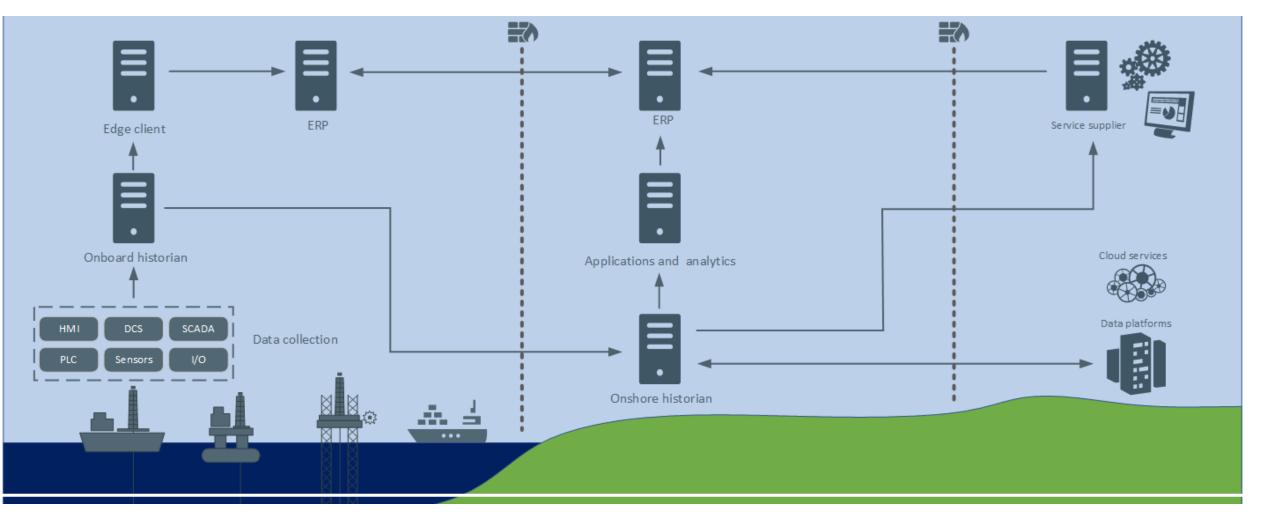
Industry software solutions Data management and analytics

Consulting and advisory services

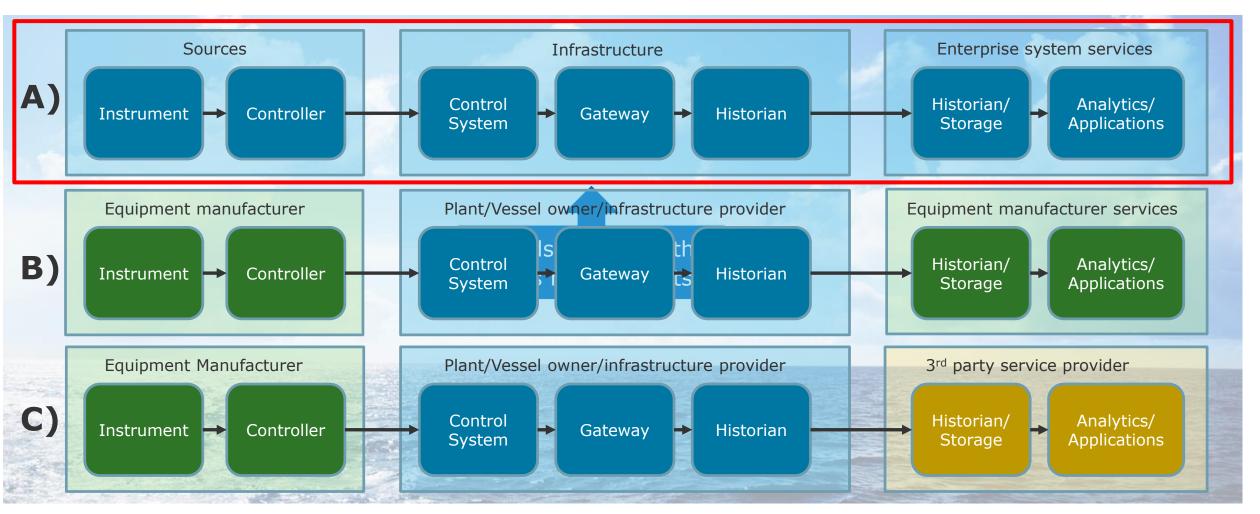
Industry data platform

Cyber security

Digital Class drivers



Complicated data chains, corporate interdependencies and technical options



DNV GL customer benefits of qualifying products like the OSIsoft PI System



Digital trust

Managing risks and opportunities of increasingly data-driven operations and finding new ways of building trust

Data quality



Data flow and sensor systems



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Data quality perspectives



Macro:

- Organizational processes and adherence to quality systems and business goals

Meso level:

- Architecture/integration
- Planning and tools

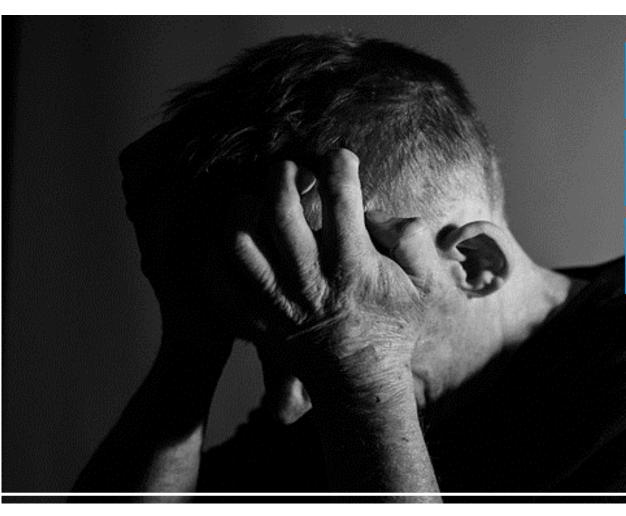
Micro: - Connectivity issues

- Dead/stale points

Data quality - Day to day issues



Data quality - Big incidents

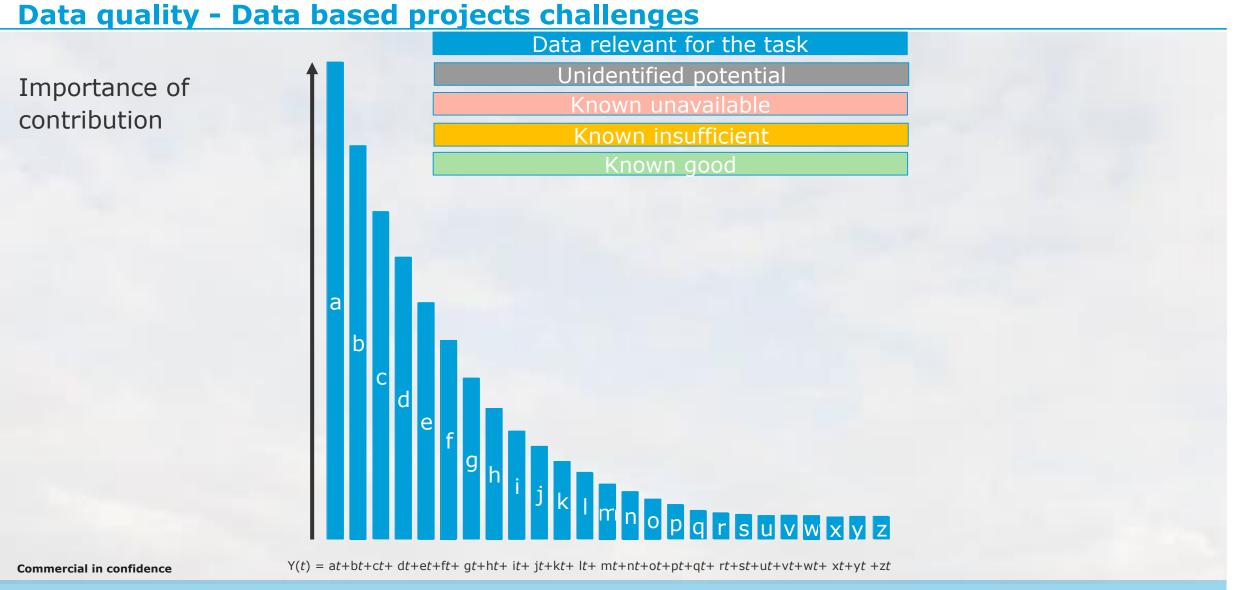


Data collection stopped working for several hours at the worst possible time

Permanent loss of years of plant data during a software upgrade

Equipment monitoring package rendered useless due to poor event quality

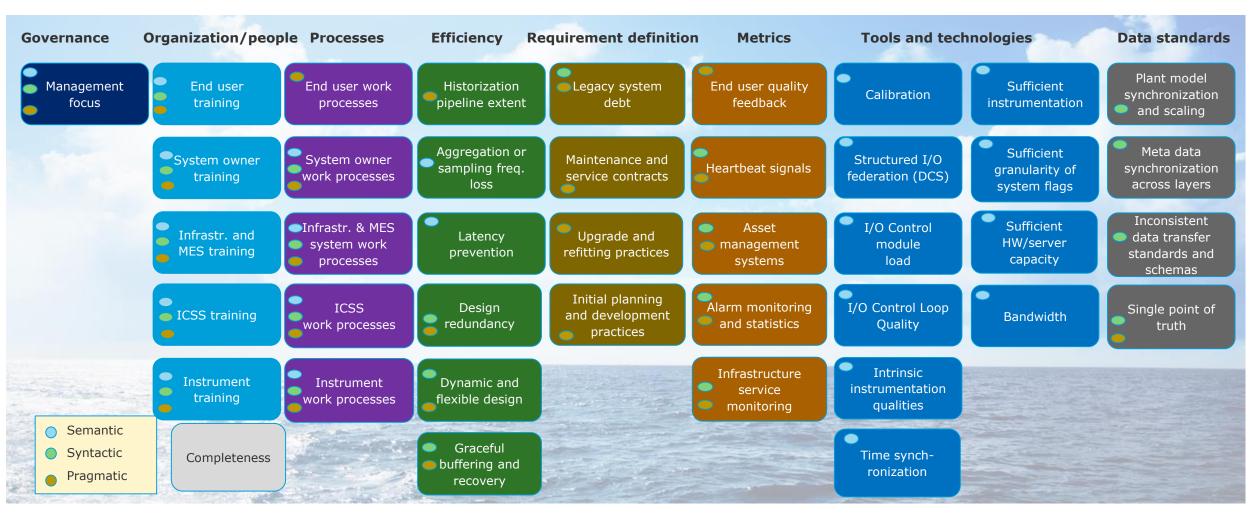
Predictive maintenance and analytics projects failed due to data quality issues



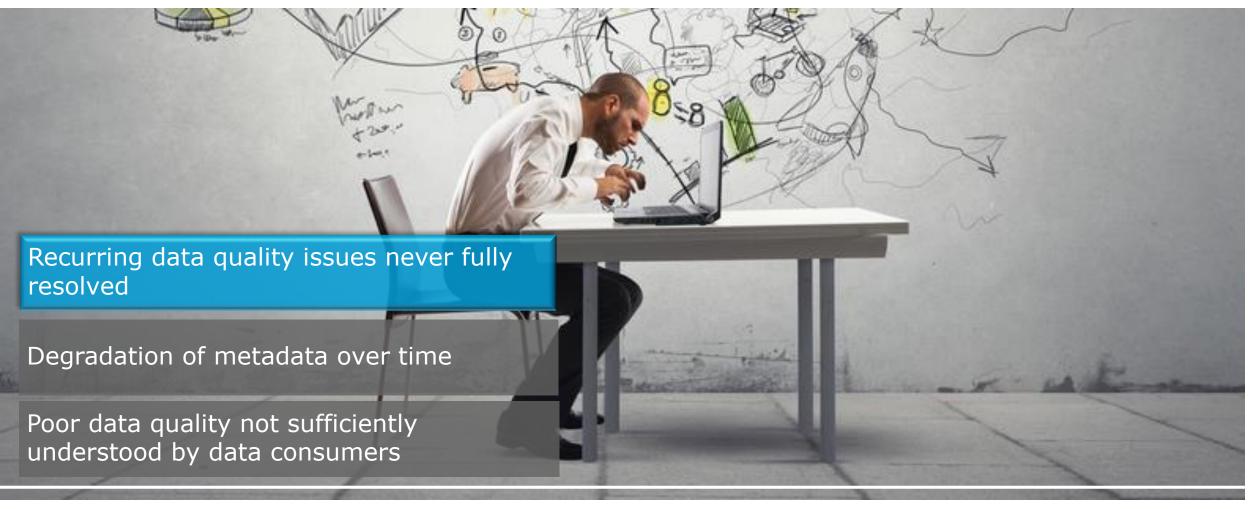
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Raise your hand: Who has started working on data quality improvement due to such issues?

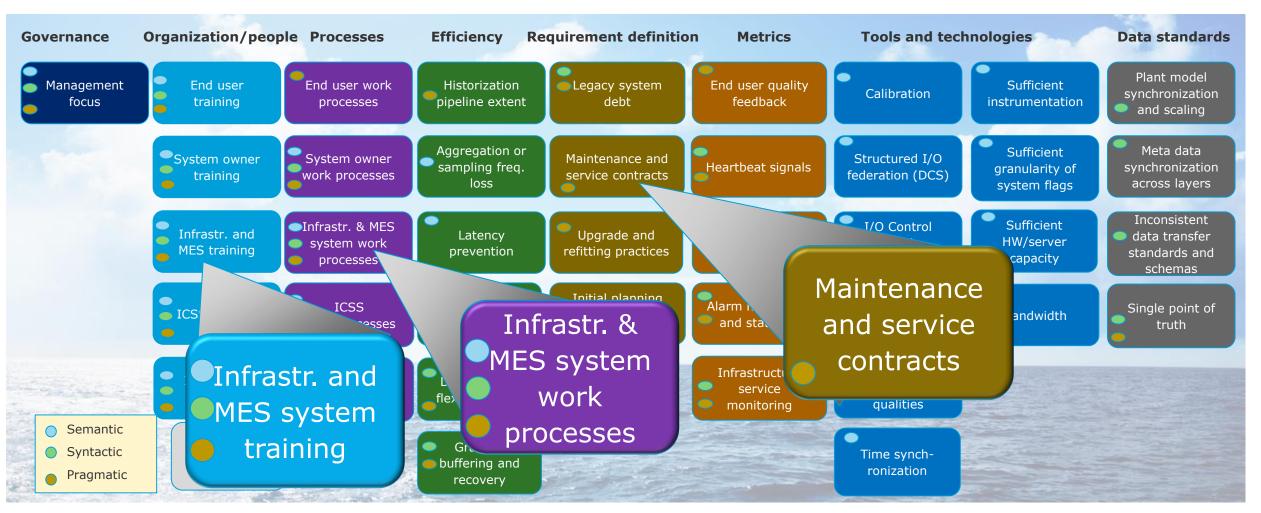
Sensor system data quality dimensions



Data quality - Day to day issues



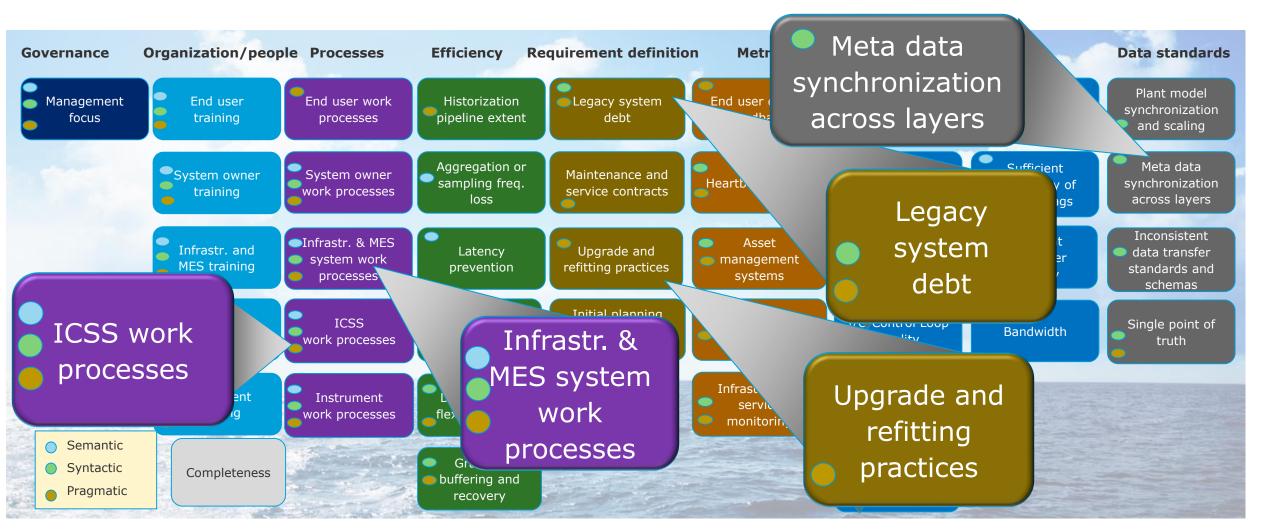
Recurring data quality issues never fully resolved



Data quality - Day to day issues



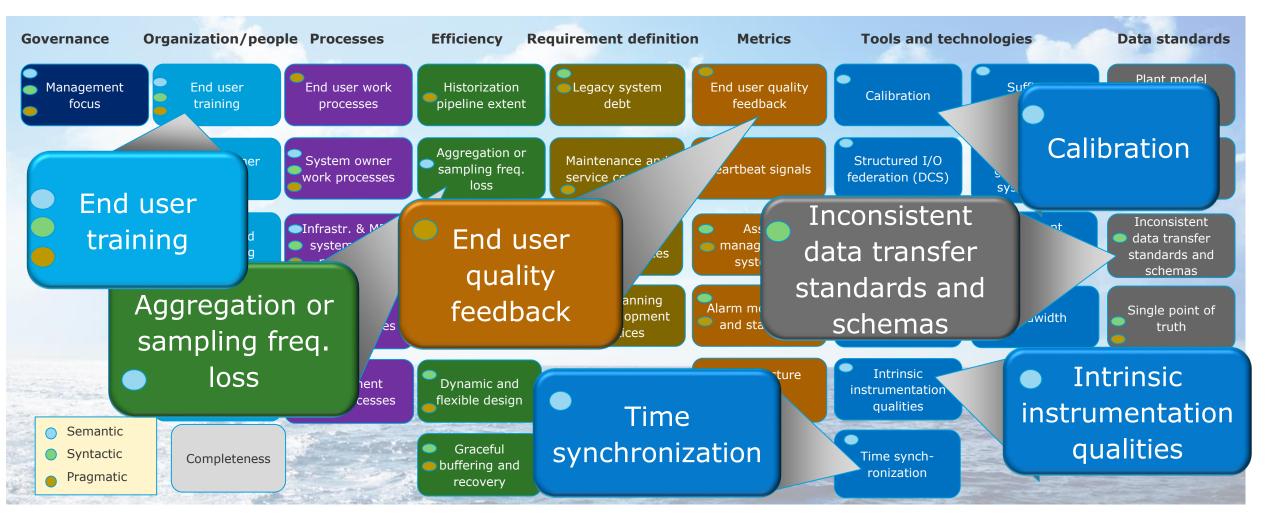
Degradation of metadata over time



Data quality - Day to day issues



Poor data quality not sufficiently understood by data consumers



Data quality - Big incidents



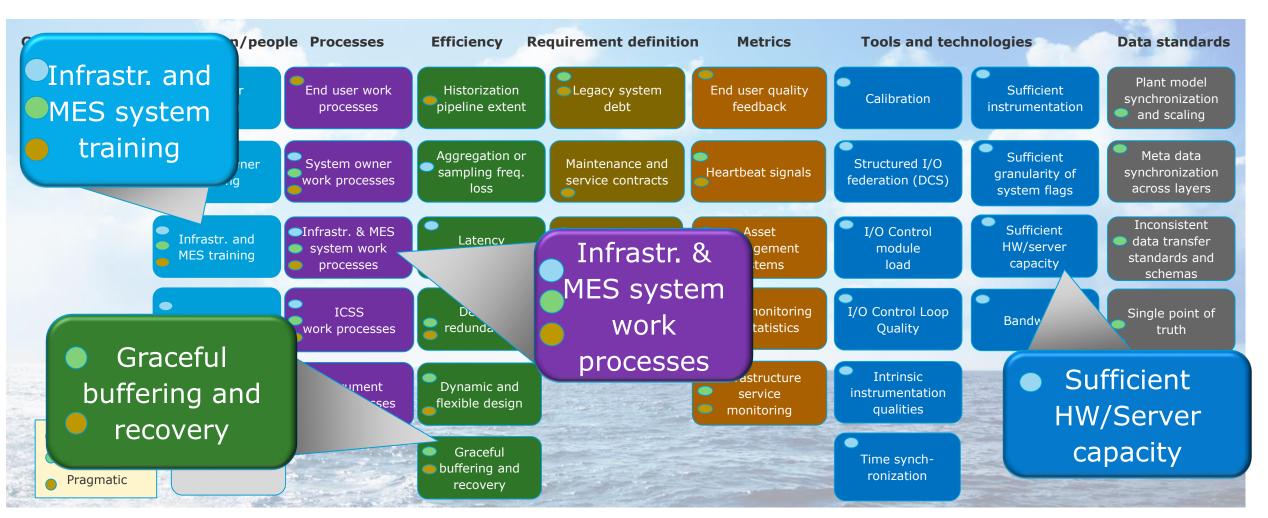
Data collection stopped working for several hours at the worst possible time.

Permanent loss of years of plant data during a software upgrade

Equipment monitoring package rendered useless due to poor event quality

Predictive maintenance and analytics projects failed due to data quality issues

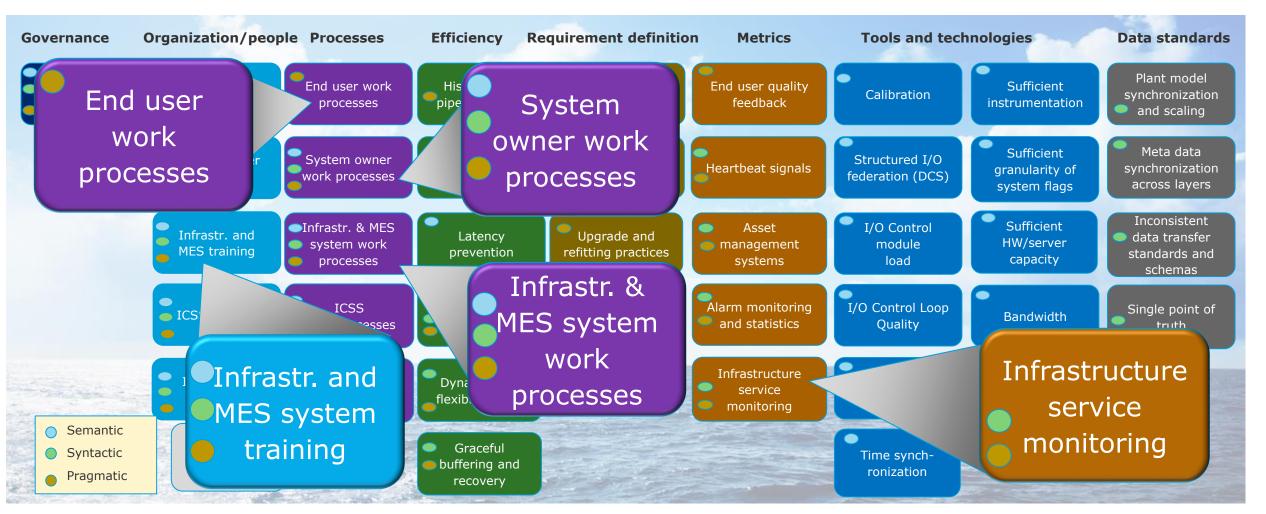
Data collection stopped working for several hours at a critical time



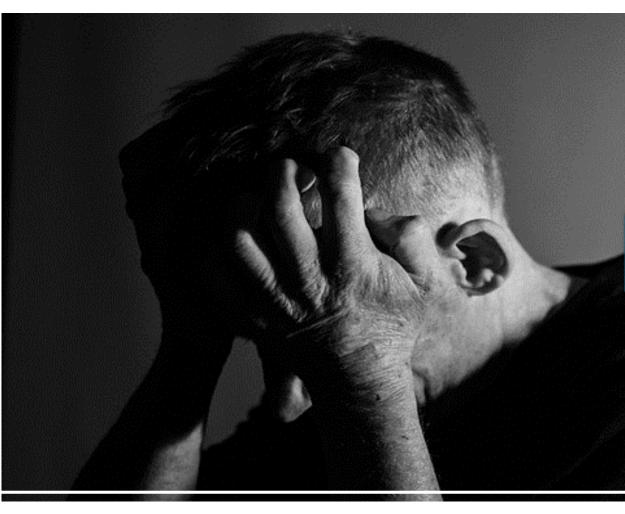
Data quality - Big incidents



Permanent loss of plant data during software upgrade



Data quality - Big incidents



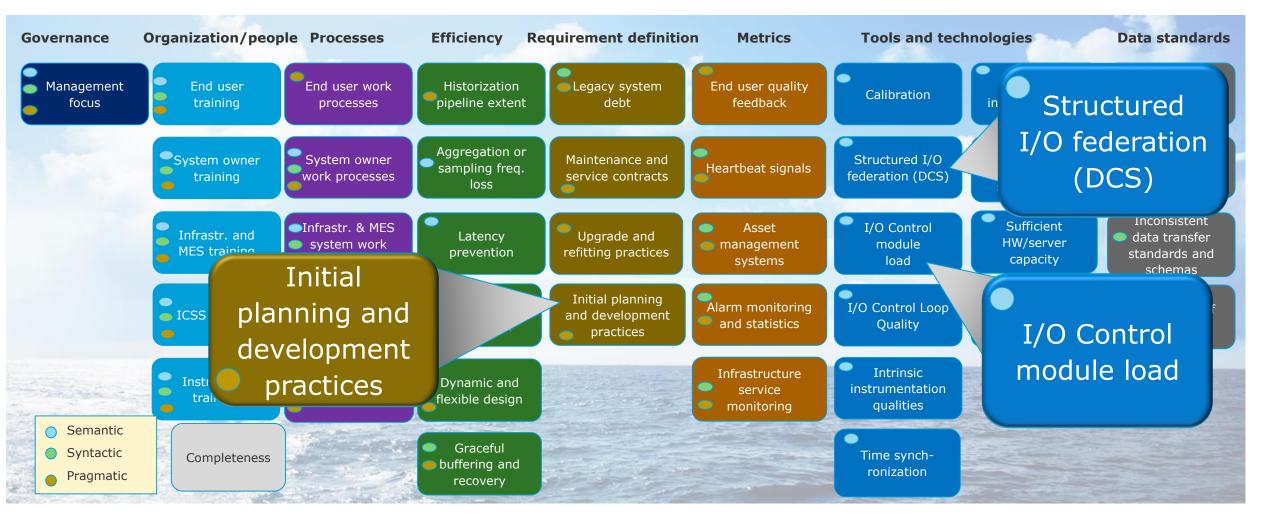
Data collection stopped working for several hours at the worst possible time

Permanent loss of years of plant data during a software upgrade

Equipment monitoring package rendered useless due to poor event quality

Predictive maintenance and analytics projects failed due to data quality issues

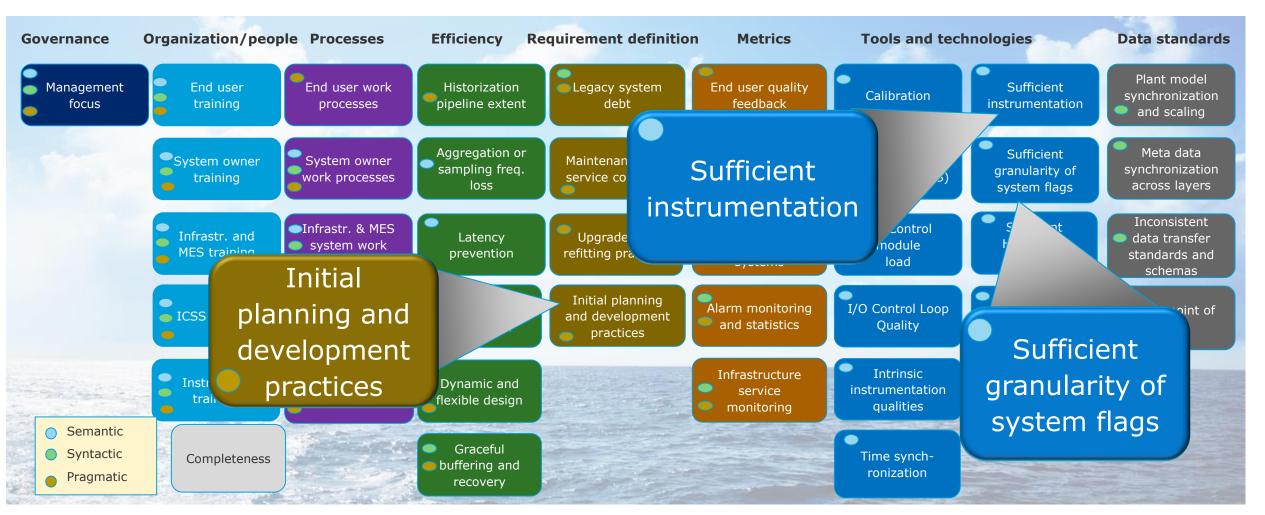
Equipment monitoring package rendered useless



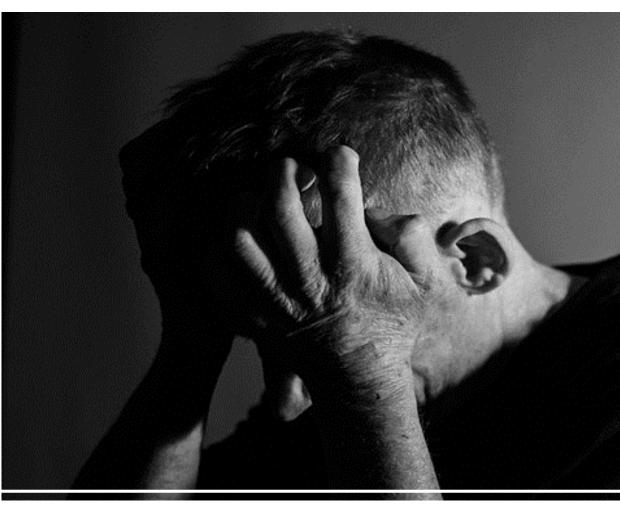
Data quality - Big incidents



Predictive analytics projects fail due to poor data quality – unavailable data



Data quality - Big incidents



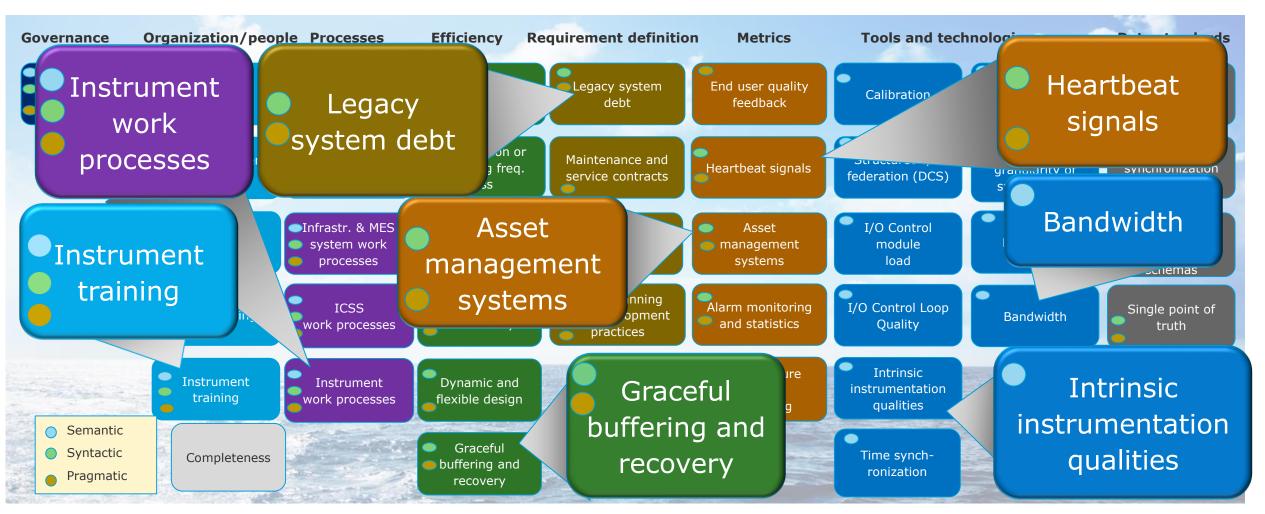
Data collection stopped working for several hours at the worst possible time

Permanent loss of years of plant data during a software upgrade

Equipment monitoring package rendered useless due to poor event quality

Predictive maintenance and analytics projects failed due to data quality issues - **insufficient**

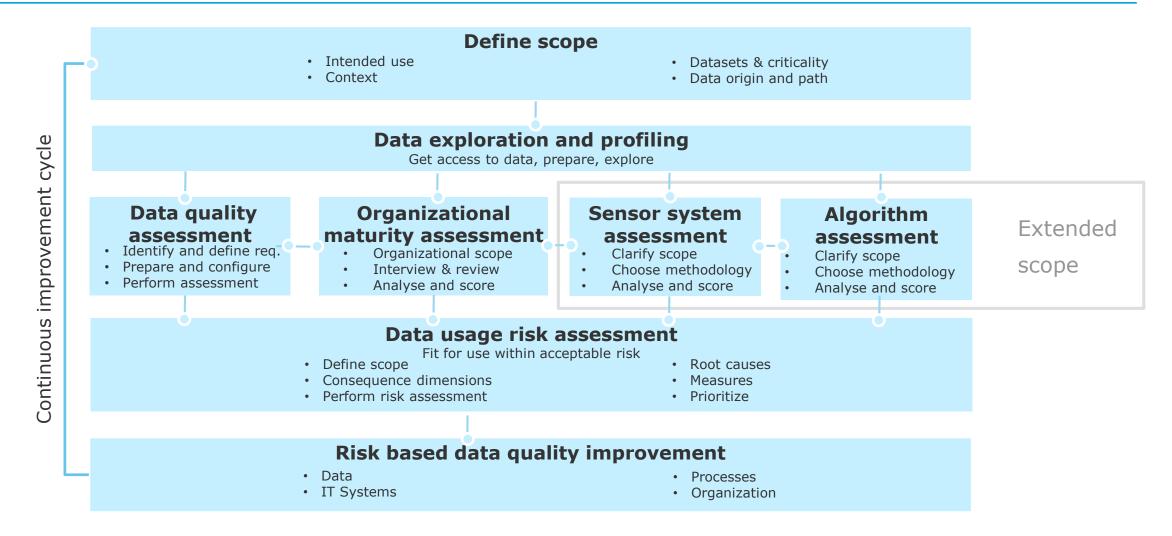
Predictive analytics projects fail due to poor data quality – insufficient data



- Establish data requirements early in new builds
- Include data quality KPIs in service contracts
- Facilitate cross discipline communication
- Have good data quality feedback systems
- Data is an asset. Think condition monitoring



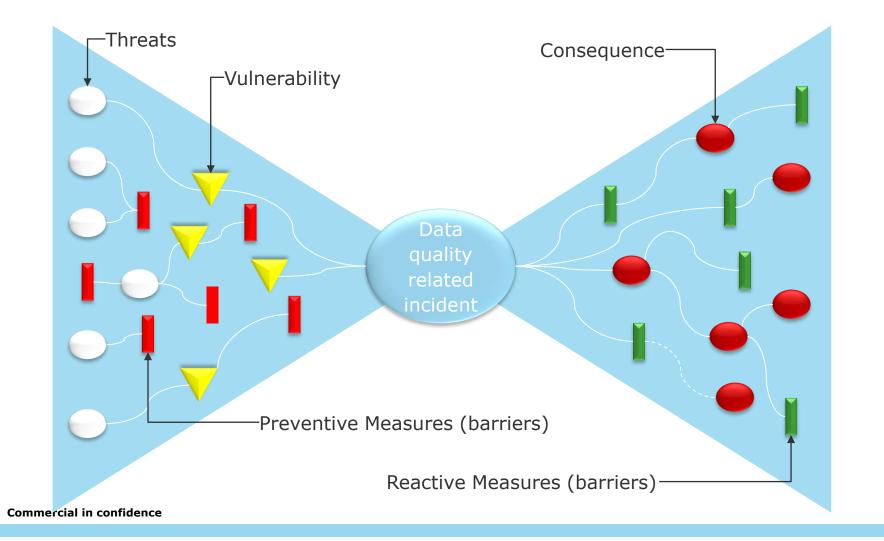
Specific recommendations – based on advisory services



Organizational Maturity Assessment

Maturity level	Governance	Organization and people	Processes	Process Efficiency	Requirement definition	Metrics and dimensions	Architecture, tools and technologies	Data standards
LEVEL 5 - Optimized	Data management policies governs and drives improvements	Data management board oversees improvement activities	Processes for continuous improvement in place	Processes provides feed- back and feed- forward to support continuous improvement	Baseline established and improvements measured according to requirements	Metrics defines baseline to support continuous improvement	Tools support policy driven continuous improvement cycle	Standard compliance and domain models are subject to continuous improvement
LEVEL 4 – Managed	Policies defined in relation to business objectives	Skillset extended to include risk analysis of quality issues aligned with business objectives	Processes for impact analysis and risk mgmt. in place	Monitoring is performed across enterprise and published as KPI's and trends	Requirements are linked to business impacts	Metrics are linked to business impacts and risk analysis	Tools are driven by business objectives and include support for root cause analysis and risk mgmt.	Standards are used actively to reduce risk for critical business operations
LEVEL 3 – Defined	Policies defined at enterprise level	Roles and required skills defined at enterprise level	Processes are defined and impemented consistently across enterprise	Defined metrics are monitored in advance of business impact	Requirements defined and communicated at enterprise level	Framework for metrics and dimensions defined at enterprise level	Archivecture in place at enterprise level supporting full stack data management	Standards, domain models and semantics used at enterprise level
LEVEL 2 – Repeatable	Local initiatives address the requirement for policies	Locally defined roles and some basic skills	Best practices in place but not used consistently	Generic metrics are monitored at point of impact	Local O initiatives define requirements	Metrics are reused locally in projects	Tools and technologies used consistently in selected projects	Industry standard and domain models used selectively across projects
LEVEL 1 - Initial	Only ad-hoc or temporal policies in place	No formally defined roles or skillset	Ad-hoc or reactive responses to quality issues	No baseline and no monitoring of quality issues	Re-engineering used to derive requirements	Project specific metrics	Tools are used ad-hoc per project	Ad-hoc and inconsistent use o standards
Objectives,	Policy, Culture, Awareness, Risks, Capabilities to handle DQ issues	Organization, roles, responsibilities, authority, skillsets	Structured and vetted ways of handling and preventing DQ issues	Measure, monitor and use metrics to mitigate DQ issues	DQ Requirements defined, communicated and acted upon	DQ metrics defined, setup, measured and monitored	DQ Tools for processing, analysing and correcting DQ issues with data assets	Use available standards, models ontologies and taxonomies – a corporate «DQ language»

Risk based continuous improvement, prevent and fix



SELF-ASSESSMENT



Data quality improvement

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CHALLENGES

- Poor data quality negatively impacts business
- Efforts to achieve structured data quality improvement become either too ad-hoc or too generic

SOLUTION

- Relate data quality issues to familiar concepts and categories
- Target improvement efforts in dedicated areas

BENEFITS

- Increased value from data
- Alignment with digital class requirements

Investing in data quality is a prerequisite for successful data science.



Per Myrseth, Manager of Data management and analytics, DNV GL

Speaker



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