

NOVEMBER 2022

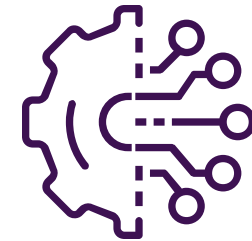
Asset Maintenance From Eye to AI

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AVEVA

Agenda

- The eye test
- Asset maintenance background
- Usage based maintenance
- CMMS Data Ingress to PI
- Condition Monitoring
- Artificial Intelligence / Predictive Analytics
- PI Data Egress to CMMS





The Eye Test

AVEVA

The Eye Test

Which asset requires maintenance?



Sometimes it's obvious

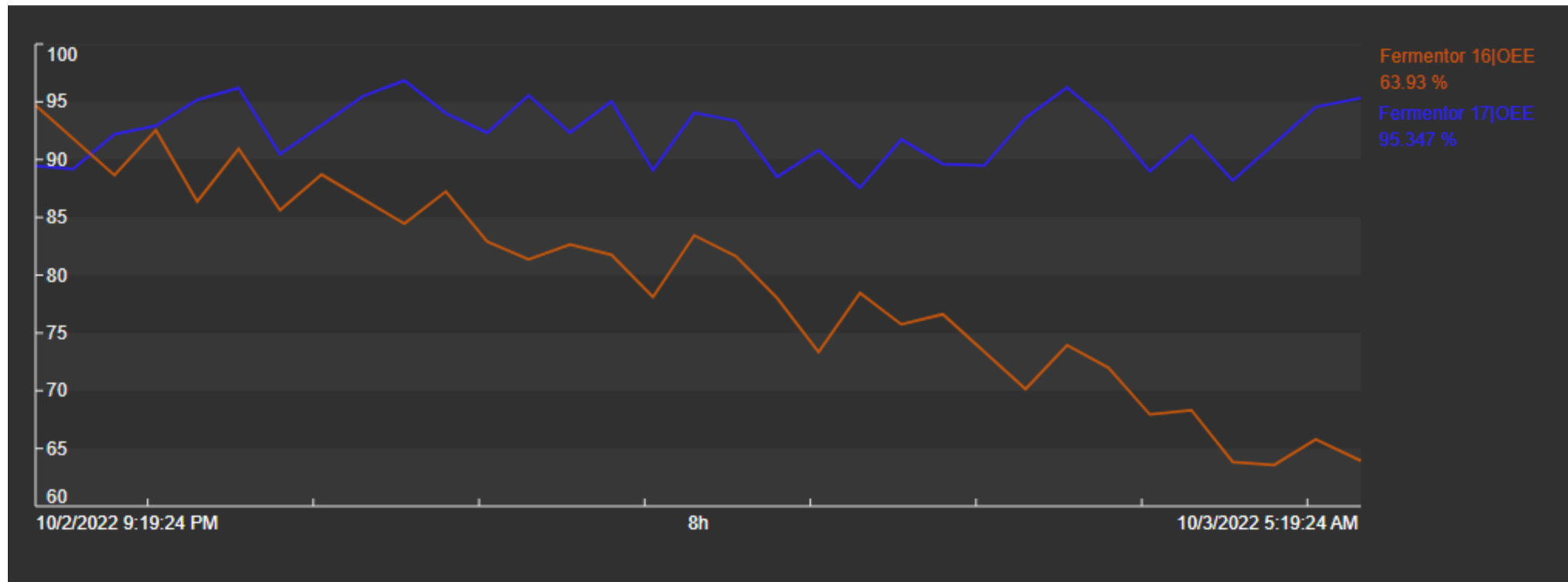


The Eye Test IRL

If only it were that easy...

The Eye Test – digital version

Which asset requires maintenance?

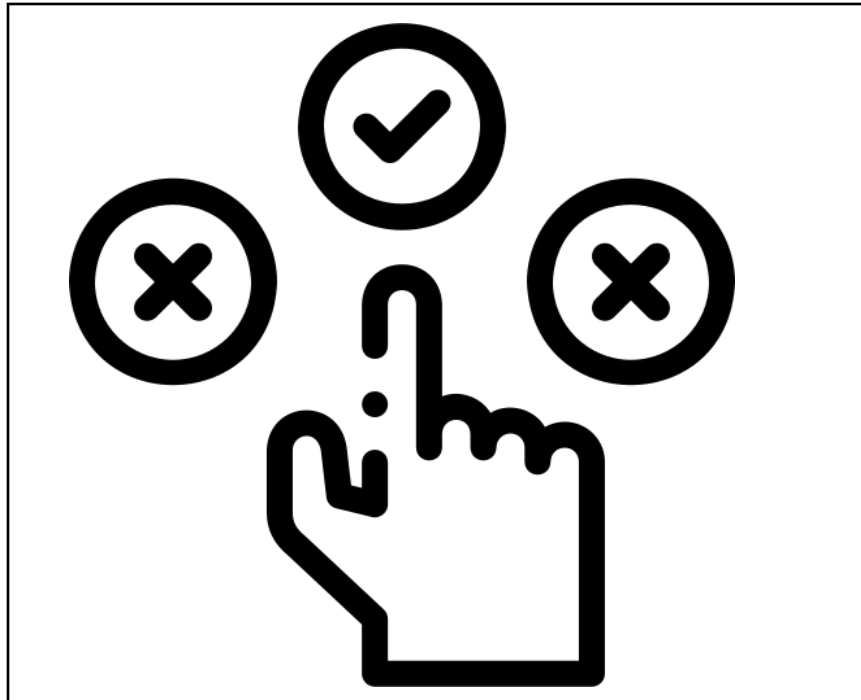




Golden rules of asset maintenance

- Clearly understand the benefit of performing maintenance
 - It must make financial sense and is technically feasible
- Maintenance is perceived as a value rather than a cost
- Understand the impact of equipment failures
 - Operational, Safety, Environment, Brand
- Understanding failure to ensure successful proactive maintenance

The Intelligent Maintenance Plan



**'Right Work at
the Right Time,
Done the Right Way,
The First time'**



(Time Based Maintenance)

- 100% of work (PM) time/usage based
- Corrective Maintenance

The Plan

**Maintenance
Methodology**

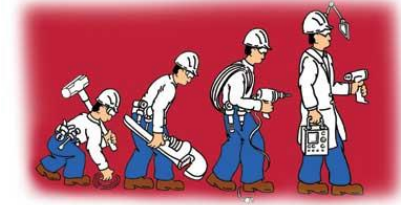


(Condition Based/Predictive Maintenance)

- Preventive, Predictive, Prescriptive
- On Condition Task (Condition Monitoring)
- Scheduled Restoration
- Scheduled Discard
- Failure Finding
- No Scheduled Maintenance

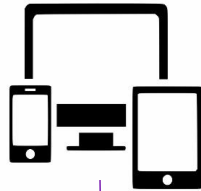
AVEVA

The past, present and future of Maintenance



Maintenance in Industry 2.0

- Pen & paper
- Basic digital platforms
 - Spreadsheets
- Formalized procedures
- No centralized system
- Mtce is primary reactive with some preventive mtce
- Limited Mtce technology



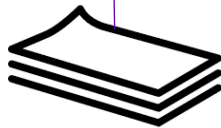
Maintenance in Industry 4.0

- Relies on combination of different digital systems
- CMMS and IIoT
- Formalized procedures with centralized system
- Mtce is primarily predictive with a balance of reactive and preventive
- Risk Based approach
- RCM, FMEA ,TPM is well established
- Mtce technology fully integrated into all operations



Maintenance in Industry 1.0

- Loosely organized
- Pen & paper
- Not centralized
- No mtce system
- Mtce Routine & Reactive
- No knowledge transfer
- No Mtce technology



Maintenance in Industry 3.0

- Primarily digital, cloud and mobile based
- Formalized procedures
- Centralized Mtce system
- Mtce primarily preventive includes reactive and predictive mtce as well
- Mtce technology used extensively
- RCM, FMEA, TPM encouraged

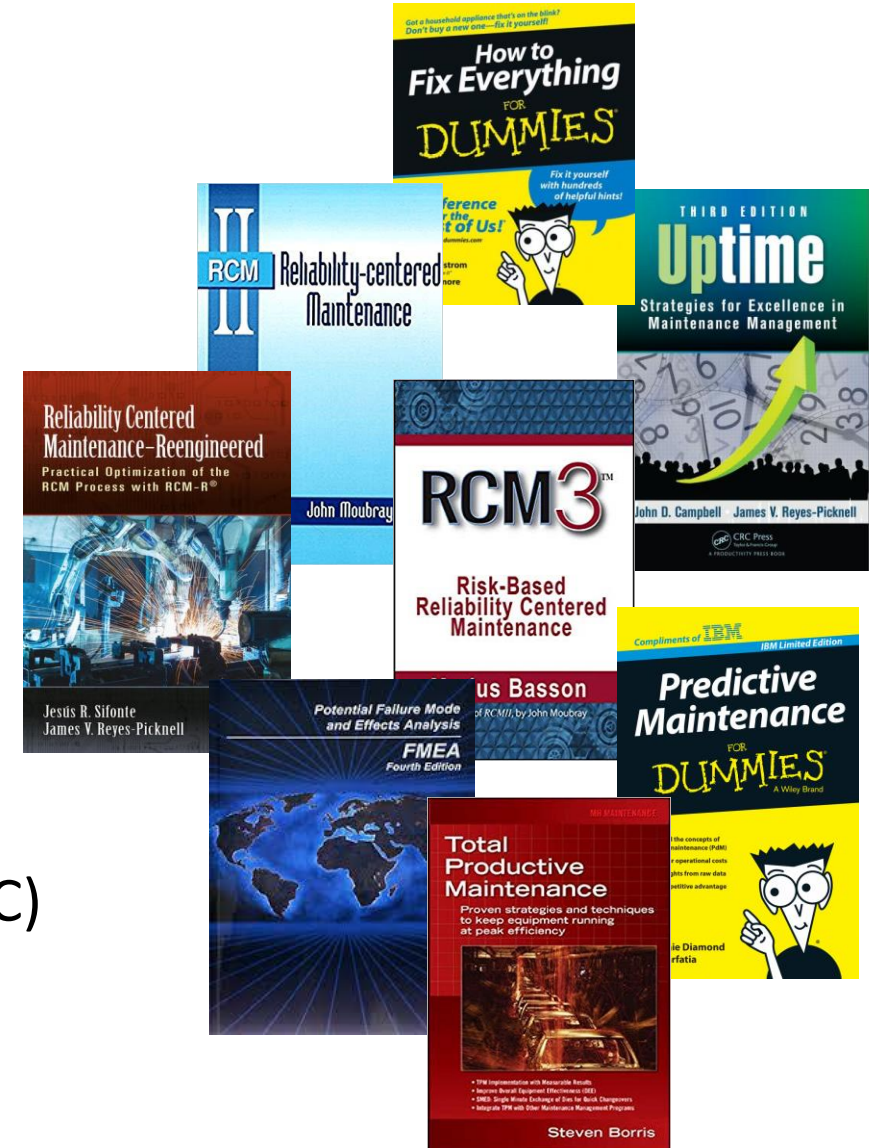


Maintenance in Industry 5.0

- Human-centric approach
- Software, hardware and people work in harmony
- Customizable platforms, integration options that fit company's operations
- IoT to IoE (Internet of Everything)
- Human-machine interaction, decision support systems, collaborative robots
- AI, ML, simulation, Digital twins, augmented, virtual or mixed
- Mtce is primarily prescriptive and predictive
- Cognitive Cyber-physical Systems

Common Maintenance Strategies

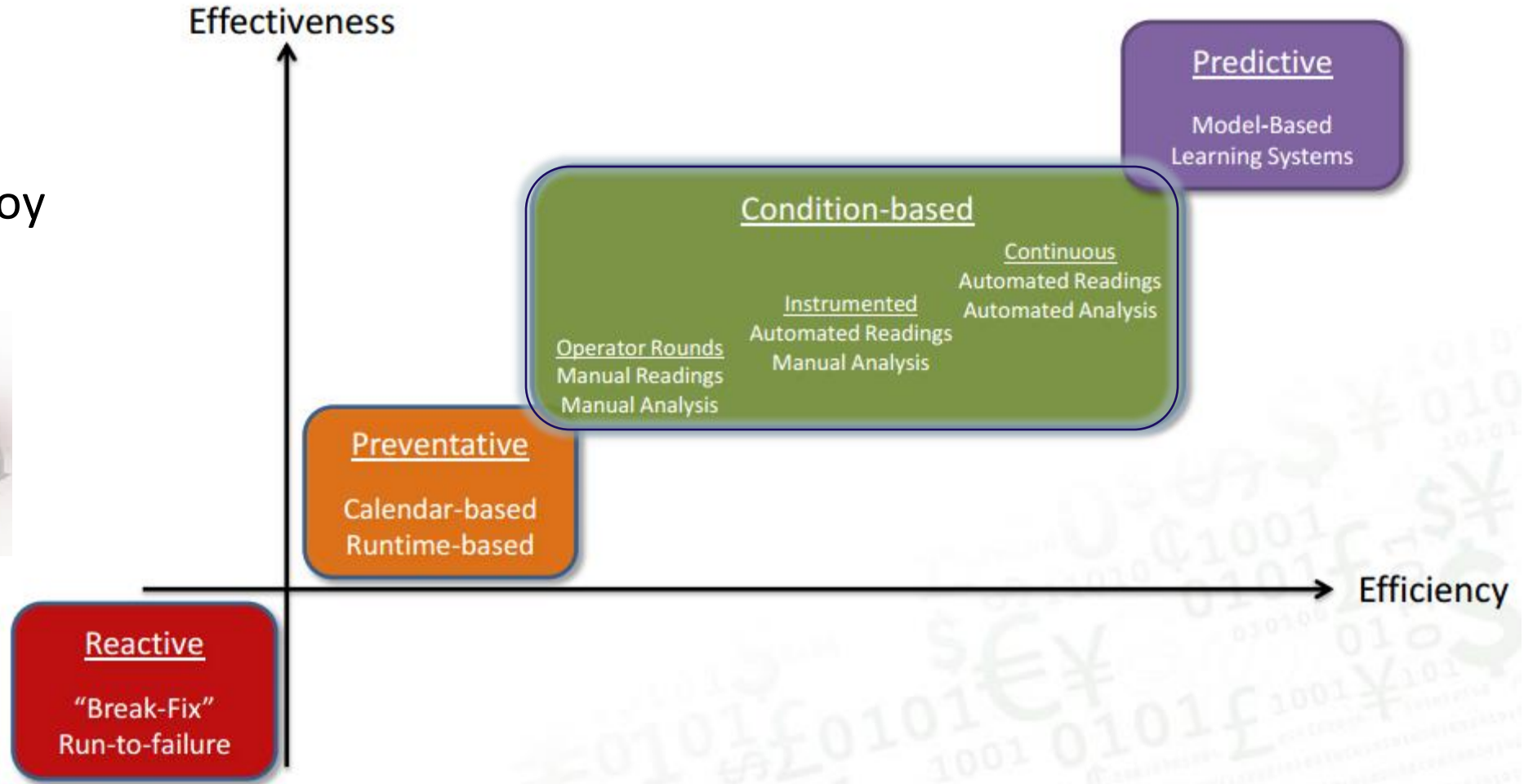
- Reactive Maintenance
- Corrective Maintenance
- Preventive Maintenance (PM)
- Predictive Maintenance (PdM)
- Condition Based Maintenance (CBM)
- Prescriptive Maintenance (RxM)
- Reliability Centered Maintenance (RCM)
- Failure Mode, Effects and Criticality Analysis (FMEAC)
 - Failure Mode and Effects Analysis (FMEA)
- Total Productive Maintenance (TPM)



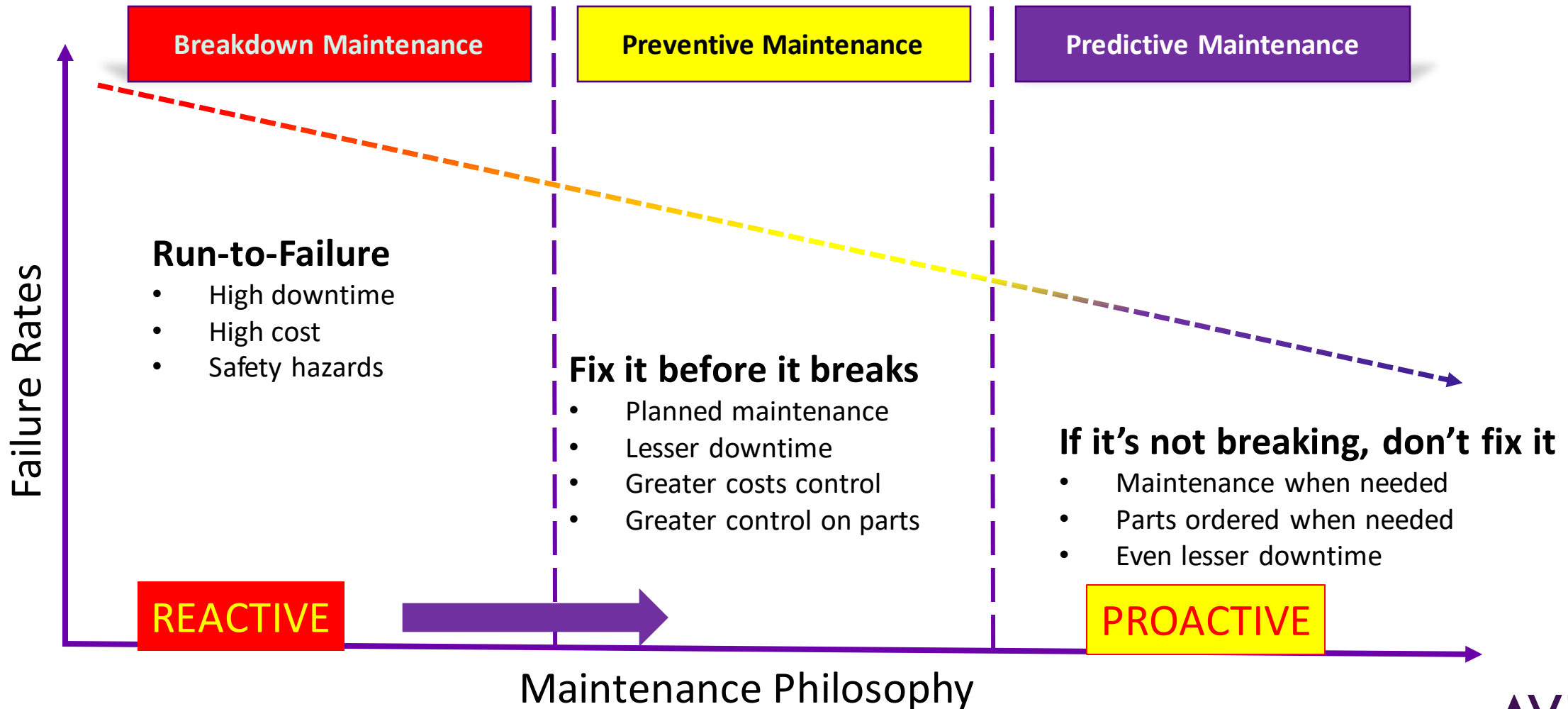
The types of traditional Maintenance

CBM:

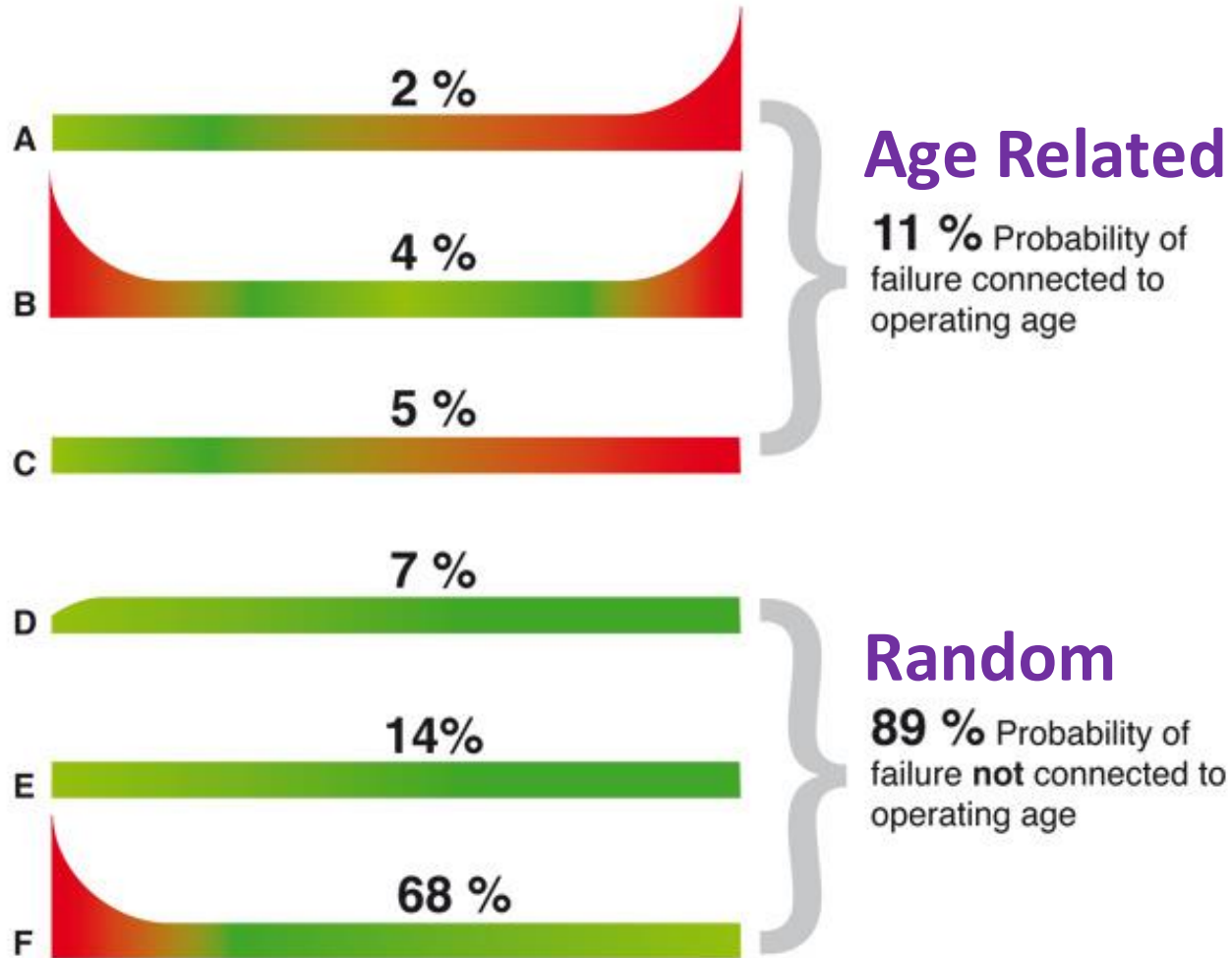
- Simple to deploy
- Greatest Value



Moving from Reactive to Proactive Maintenance



Failure Curves – Age related vs Random



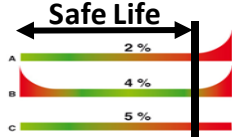
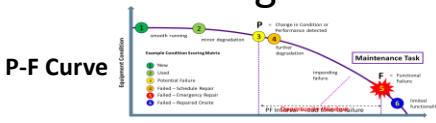
Source: RCM II by John Moubray, Industrial Press Inc, 1992

- Premature random failures
- Often after Human Intervention

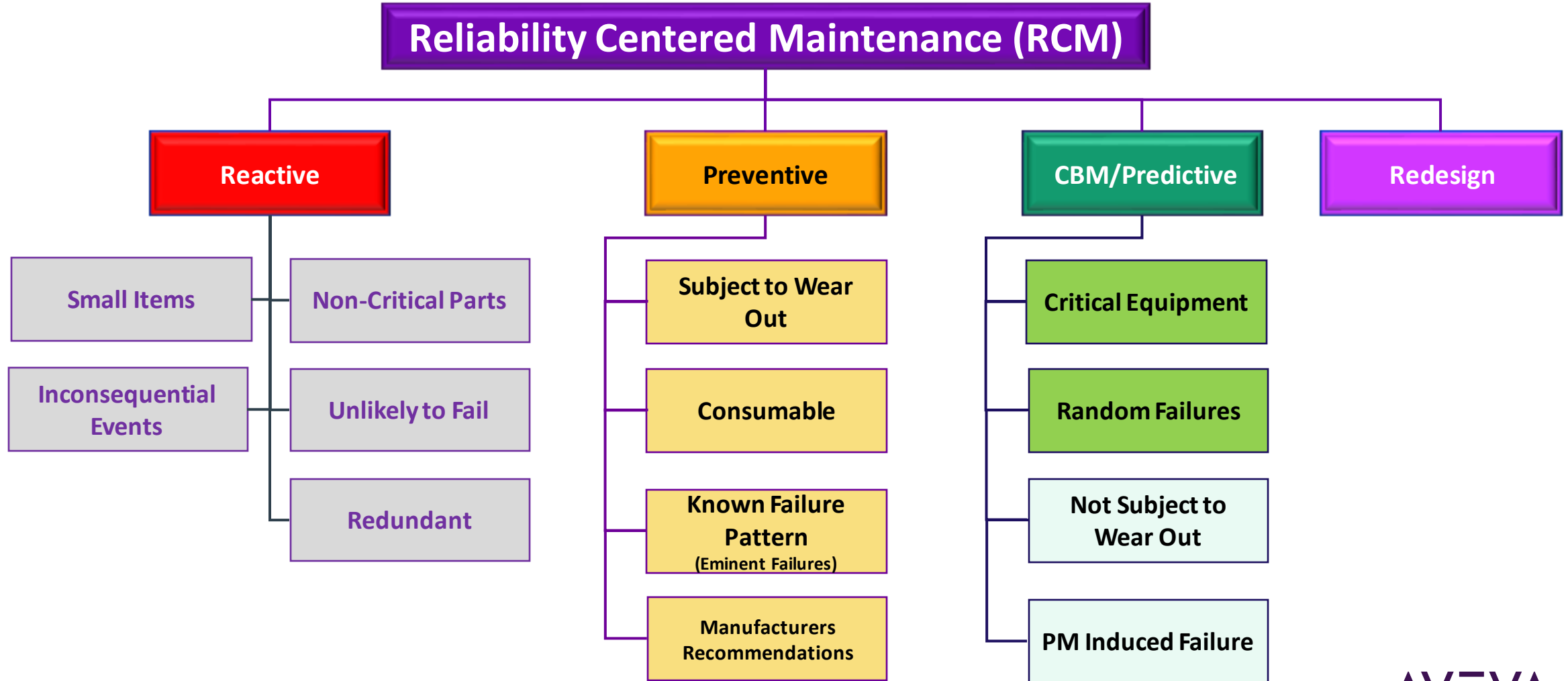
Simple method to determine maintenance strategy

4 Failure Classifications

- Age Related
- Random
- Sudden
- Gradual

	Sudden	Gradual
Age Related 11%	Time/Usage based PM at “Safe Life Interval” 	Time/Usage based PM or Condition based Maintenance (CBM)
Random 89%	Run-to-Failure with mitigation plans such as <i>redundancy</i> or <i>readily available spares</i>	Condition based Maintenance (CBM) to detect the degradation 
Aveva PI System / Aveva Predictive Analytics		

Reliability Centered Maintenance Approach





Usage Based Maintenance



Usage based maintenance

AKA Read the Manual

- Many manufacturers give recommendations for how often to maintain equipment
- Two types
 - Time based
 - Usage based
- Finding the # of run hours is the hard part



Use Asset Analytics to calculate and trigger

Back

Check In

Refresh

New Element

Search Elements

HOU FCC Pump 2

GeneralChild ElementsAttributesPortsAnalysesNotification RulesVersion

Backfilling

PumpCavitationEvent

PumpRunHours

RunHoursSinceMaint

Name:RunHoursSinceMaint

Description:

Categories:

Analysis Type:☒ Expression☐ Rollup☐ Event Frame Generation☐ SQC

Add a new variable

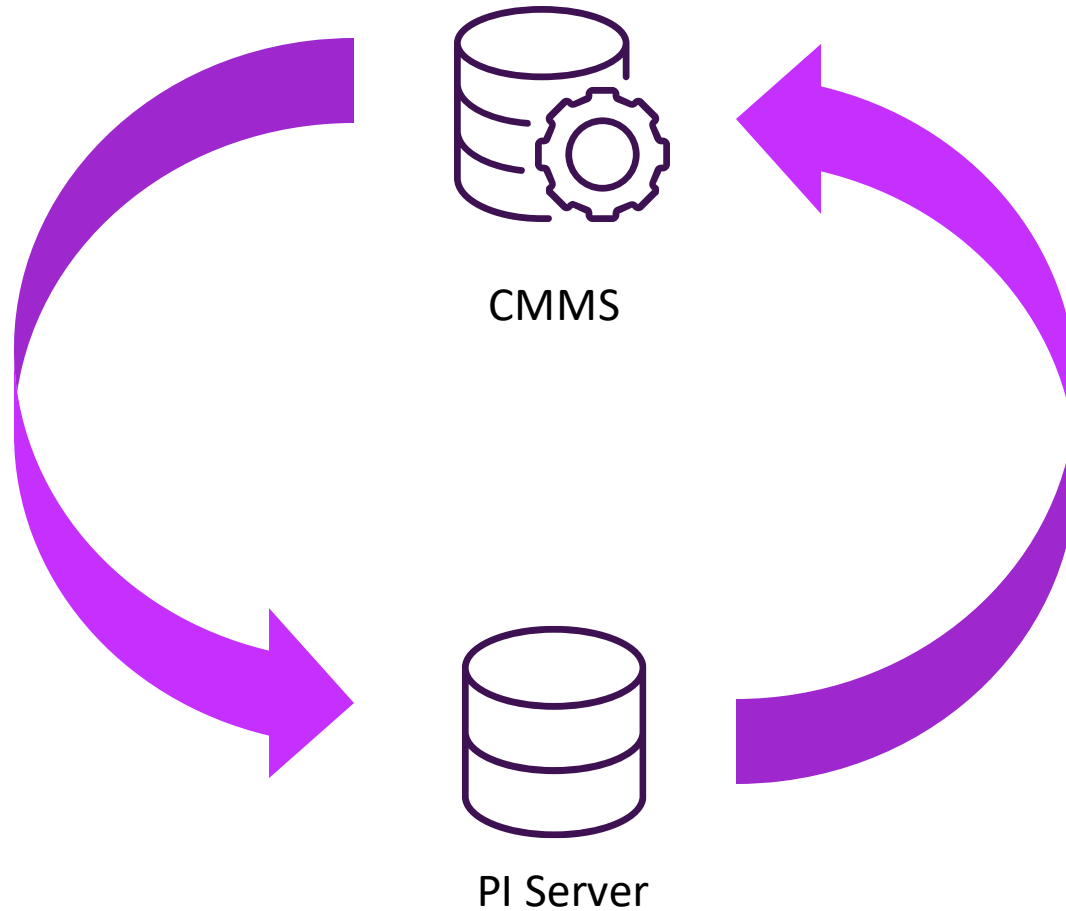
Name	Expression	Value at Evaluation	Value at Last Trigger	Output Attribute	
HoursRunSinceMaint	TimeEq('Status', 'Date - Last Service', '*', "Running")/(60*60)	9256	9256	Map	⊗
MaintenanceNeeded	HoursRunSinceMaint > 'Run Hours Before Service'	True	True	Map	⊗

Integration with Maintenance System



Maintenance Systems and the AVEVA™ PI System™

Digital collaboration



Maintenance data ingress

Common approaches

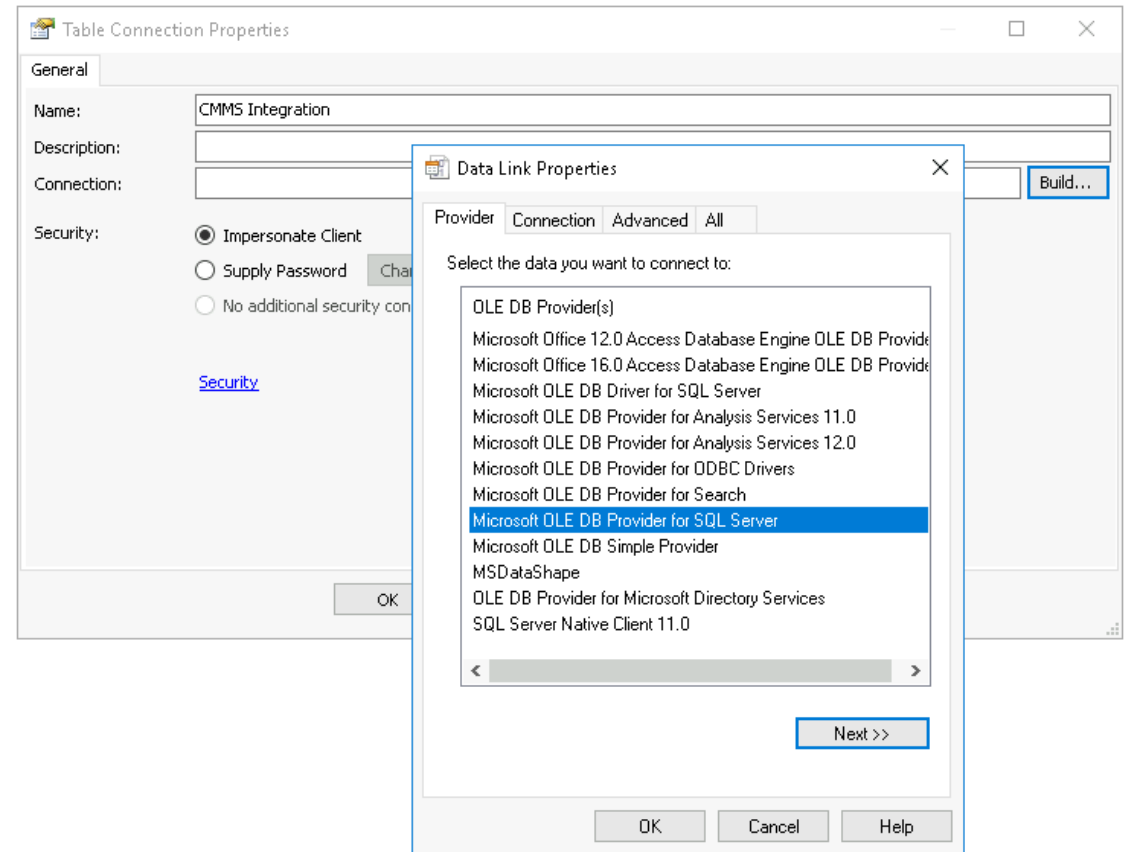
- Manually write in the date to your AF Attribute
- Use Adapter or PI Interface for RDBMS (relational database management system)
 - Stores data as PI Tags
- Use Asset Framework ODBC table connections and attribute table lookups

Category: Equipment Details		
	Date - Installed	1/1/2017 1:00:00 AM
	Date - Last Service	11/30/2020 9:30:00 AM

AF Table connection for CMMS

Single source of truth for maintenance data

- Using the data from the authoritative source
- For best performance
 - Use Parametrized Queries
 - Keep Tables as small as possible



AF Table lookup for CMMS

Single source of truth for maintenance data

- Import data or Link data option
- Query can be a simple “select” statement

AF Table (CMMS data)

CMMS Station Location Equipment													
General Table Define Table Version													
CMMS Station Location Equipment													
Filter													
enterprise_id	name	locationid	locationtype	city	region	GPSLatitude	GPSLongitude	LocationStatus	OilContainment	StreetName	StreetNum:	WorkArea	EQUIP_ID
3	TS7-Pickrel	TS7	TS	Rome	York	43.810562066	-79.478469565	IS	SUMP WELL	Dufferin Street	8000	South	1128
4	TS7-Pickrel	TS7E	TS	Rome	York	43.810562066	-79.478469565	IS	SUMP WELL	Dufferin Street	8000	South	1129
5	TS8-Pranha	TS8	TS	Rome	York	43.780209788	-79.542493404	IS	SUMP WELL	Century Place	1	South	1130
6	TS9-Shark	TS9	TS	Rome	York	43.804643049	-79.646716852	IS	SUMP WELL	Rutherford Road	6531	South	1131
7	TS5-Orata	TS5	TS	Vienna	York	43.839901184	-79.417405499	IS	SUMP WELL	Bayview & Highway 7	160	South	1132
8	TS6-Perch	TS6	TS	Vienna	York	43.839911818	-79.416522441	IS	SUMP WELL	Bayview & Highway 7	160	South	1133
9	TS1-Manta	TS1	TS	Athens	York	43.835947766	-79.340634285	IS	SUMP WELL	14th Avenue	3430	South	1134
10	TS2-Marin	TS2	TS	Athens	York	43.859719536	-79.257314884	IS	SUMP WELL	Hwy 48	7970	South	1135
11	TS3-Moray	TS3	TS	Athens	York	43.84824685	-79.311939307	IS	SUMP WELL	Kennedy Road	7932	South	1136
11	TS3E-Moray Exp	TS3E	TS	Athens	York	43.84824685	-79.311939307	IS	SUMP WELL	Kennedy Road	7932	South	1137
13	TS4-Sturgeon	TS4	TS	Athens	York	43.84399640	-79.3506525	IS	SUMP WELL	Addscott Court	85	South	1138
17	D59-Albacore	D59	MS	Athens	York	43.834197065	-79.340339034	IS	SUMP WELL	14th Avenue	3451	South	1071
18	D510-Beluga	D510	MS	Athens	York	43.826011769	-79.423543675	IS		Baythorn Drive	227	South	1072
19	D512-Bitterling	D512	MS	Athens	York	43.807079193	-79.419636359	IS	NONE	Morgan Avenue	30	South	1073
20	D511-Pigeon	D511	MS	Athens	York	43.819978988	-79.387792991	IS	NONE	John Street	397	South	1074
23	Concord	VM51	MS	Rome	York	43.803557108	-79.500607968	dc	NONE	Keele Street	7896	South	1077
24	King	VM58	MS	Rome	York	43.909372558	-79.525264956	IS	NONE	Malay Street	200	South	1078
25	D51-Androvy	D51	MS	Geneva	York	43.995472312	-79.458651926	IS	NONE	Edward Street	135	South	1079
26	D52-Angler	D52	MS	Geneva	York	44.011581532	-79.468867783	IS	NONE	Old Yonge Street	21	South	1080
27	D54-Arowana	D53	MS	Geneva	York	43.99582483	-79.495096173	IS	NONE	Bathurst Street	15459	South	1081
28	D54-Barjo	D54	MS	Geneva	York	43.965609354	-79.48721361	IS	NONE	Bathurst Street	14025	South	1082
29	D55-Barbel	D55	MS	Geneva	York	44.013244596	-79.446138722	IS		Bayview Avenue	15560	South	1083
30	D56-Barracuda	D56	MS	Geneva	York	43.9653851	-79.442335764	IS		Bayview Avenue	14778	South	1084
31	D57-Barreleye	D57	MS	Geneva	York	44.017594062	-79.419760533	IS	SORBEWEB	Leslie Street	15521	South	1085
32	D58-Bass	D58	MS	Geneva	York	44.011453785	-79.418300465	IS	SUMP WELL	Leslie Street	15267	South	1086
33	D5301-Bowfish	D5301	MS	Prague	Simcoe	44.40427575	-79.723714319	IS	NONE	Anne Street North	381	North	1087
34	D5302-Bullnose	D5302	MS	Prague	Simcoe	44.33228839	-79.675919854	IS		Saunders Road	169	North	1088
35	D5303-Boga	D5303	MS	Prague	Simcoe	44.355778253	-79.715372393	IS		Ferndale Drive South	202	North	1089

AF Attribute Table Lookup

Table Lookup Data Reference

Table:

CMMS Station Location Equipment

Result column:

Last Maintenance Date

Unit of Measure:

<None>

Behavior

Rule:

Select first row matching criteria

Order by:

Last Maintenance Date

ASC

Where

Column:

Horsepower (BHP)

Operator:

=

Attribute or Value:

@Amps - Motor

Add And

Add Or

Complete WHERE Clause:

[name] = '%Element%'

Table Parameters

Parameter

Value

Replacement Values

Value to return when no matching row found:

No Data

Value to return when NULL result found:

DBNull

OK

Cancel



Condition Monitoring

What is CBM?

- Maintenance strategy focused on preventing asset failures, downtime and unnecessary maintenance practices by monitoring asset health in order to determine what maintenance needs to be completed and when
- Asset Health is based on Real-time or near Real-time assessment of asset condition using different on-condition techniques
- CBM includes:
 - Condition Monitoring process
 - Corrective maintenance activities



Types of Condition-Monitoring Systems

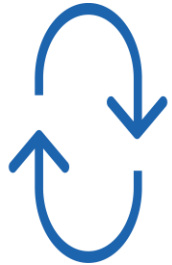
Two Types: Periodic and Continuous

Periodic Monitoring Systems



- Acquire at selected time intervals (hourly, daily, weekly, monthly patrols)
- Analysis of the collected data is generally performed in the office or laboratory
- Measurements may be on-line or offline, depending on what is being measured and the access required

Continuous Monitoring Systems



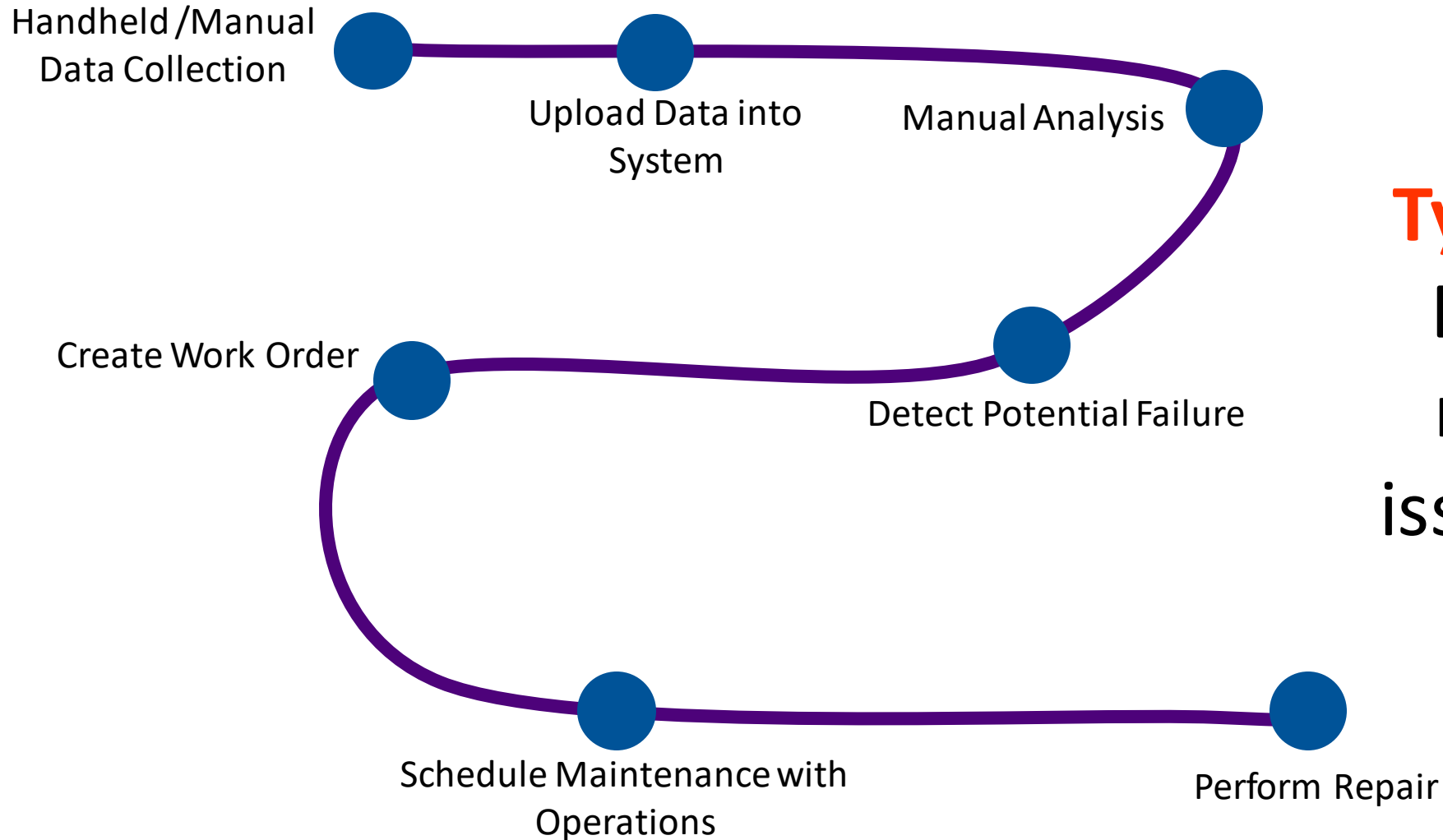
- Collect **almost always on-line** the measurement values at selected points on the equipment and this data is continuously compared with pre-established levels or criteria.
- “Set-points” are usually established for the purpose of providing automatic warnings and/or equipment shutdown, depending on the consequence of the impending failure.

Selection of Periodic or Continuous:

- Periodic: time and resource limitations for conducting periodic; on- or off-line monitoring
- Continuous: hardware and installation costs for Continuous, on-line monitoring.

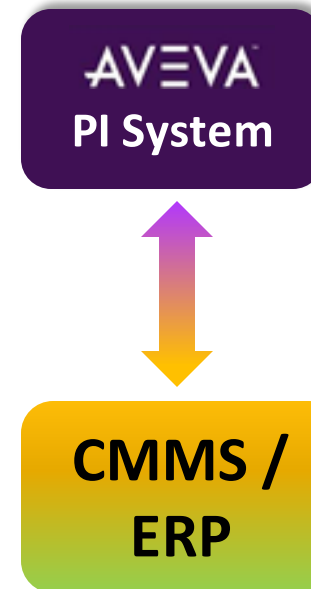
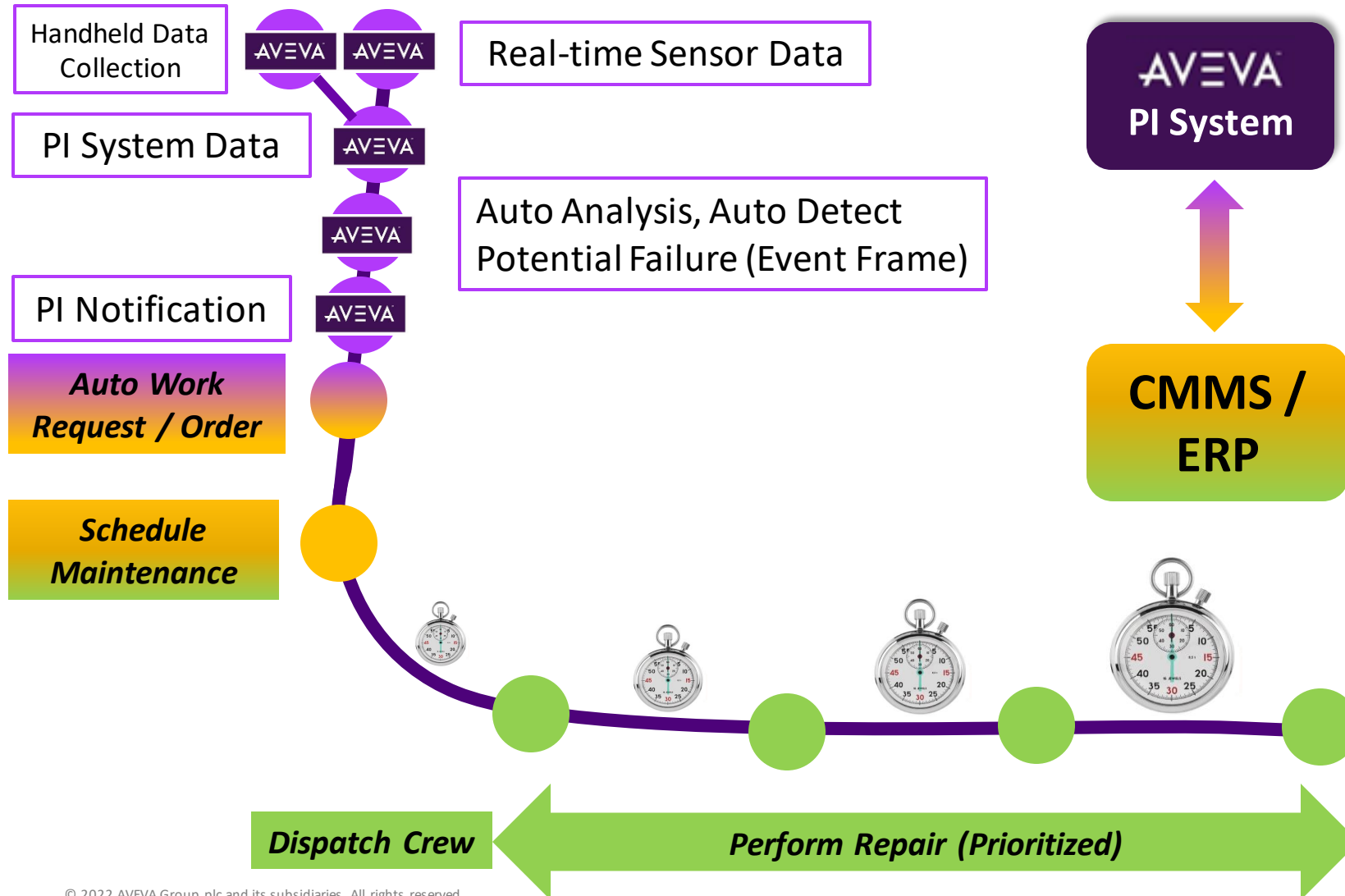


Challenge: Long Gaps Between Measurement and Action



Typical Timeline:
Days, weeks or
months before
issue is addressed

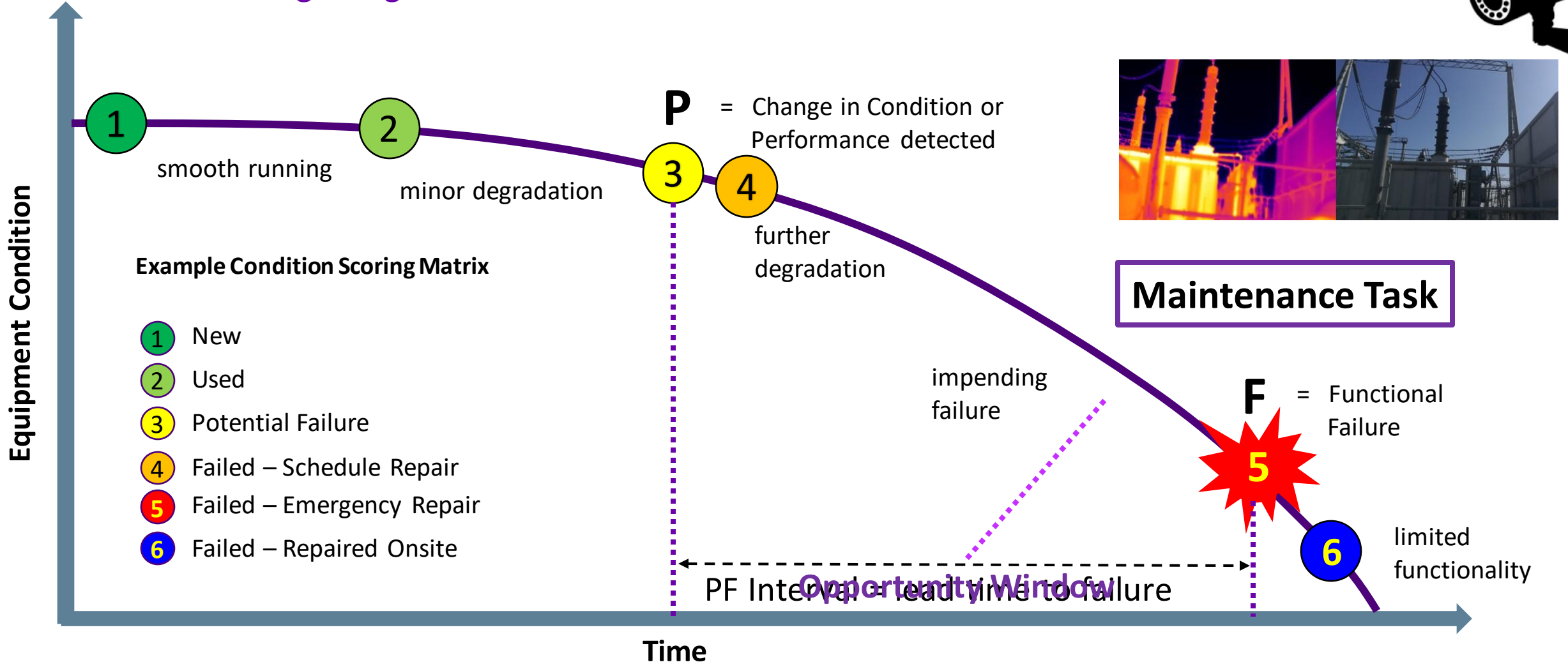
Solution: Minimize Gaps Between Measurement and Action



Typical Timeline:
Seconds to Minutes
before made aware
of issue

Condition Monitoring / Incipient Failure Detection

Condition Scoring Using Potential Failure "PF" Curve



Example PI Vision Report - Condition Scoring

Transformer Fleet Oil Condition Report

Condition Codes

1 – Normal

2 – Warning

3 – Alarm

4 – Critical

Condition Score	PI Notification	CMMS Alert	CMMS Work Order	Health Score
Code 1	N	N	N	N
Code 2	Y	Y	N	Y
Code 3	Y	Y	Y	Y
Code 4	Y	Y	Y	Y

TXTS Fleet Oil Condition

TS Transformer Fleet - Oil Condition Report

Source: TOA4 Home

	DGA	DGA Test Date	Moisture	Moisture Test Date	Fluid Quality	FQ Test Date
TS4:Sturgeon:T1	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	1	2019-07-24 12:00:00 AM
TS4:Sturgeon:T2	1	2019-10-23 11:43:00 AM	1	2019-10-23 11:43:00 AM	1	2019-07-24 12:00:00 AM
TS5:Orata:T1	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	2	2019-07-24 12:00:00 AM
TS5:Orata:T2	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	1	2019-07-24 12:00:00 AM
TS7:Pickerel:T1	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	2	2019-07-25 12:00:00 AM
TS7:Pickerel:T2	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	2	2019-07-25 12:00:00 AM
TS7E:Pickerel Exp:T3	1	2019-10-23 11:43:00 AM	1	2019-10-23 11:43:00 AM	2	2019-07-25 12:00:00 AM
TS7E:Pickerel Exp:T4	1	2019-10-23 11:43:00 AM	1	2019-10-23 11:43:00 AM	1	2019-07-25 12:00:00 AM
TS8:Piranha:T1	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	1	2019-07-25 12:00:00 AM
TS8:Piranha:T2	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	1	2019-07-25 12:00:00 AM
TS9:Shark:T1	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	1	2019-07-25 12:00:00 AM
TS9:Shark:T2	1	2020-02-24 7:30:00 PM	1	2020-02-24 7:30:00 PM	1	2019-07-25 12:00:00 AM

Condition Scoring

1 = Normal

2 = Warning

3 = Alarm

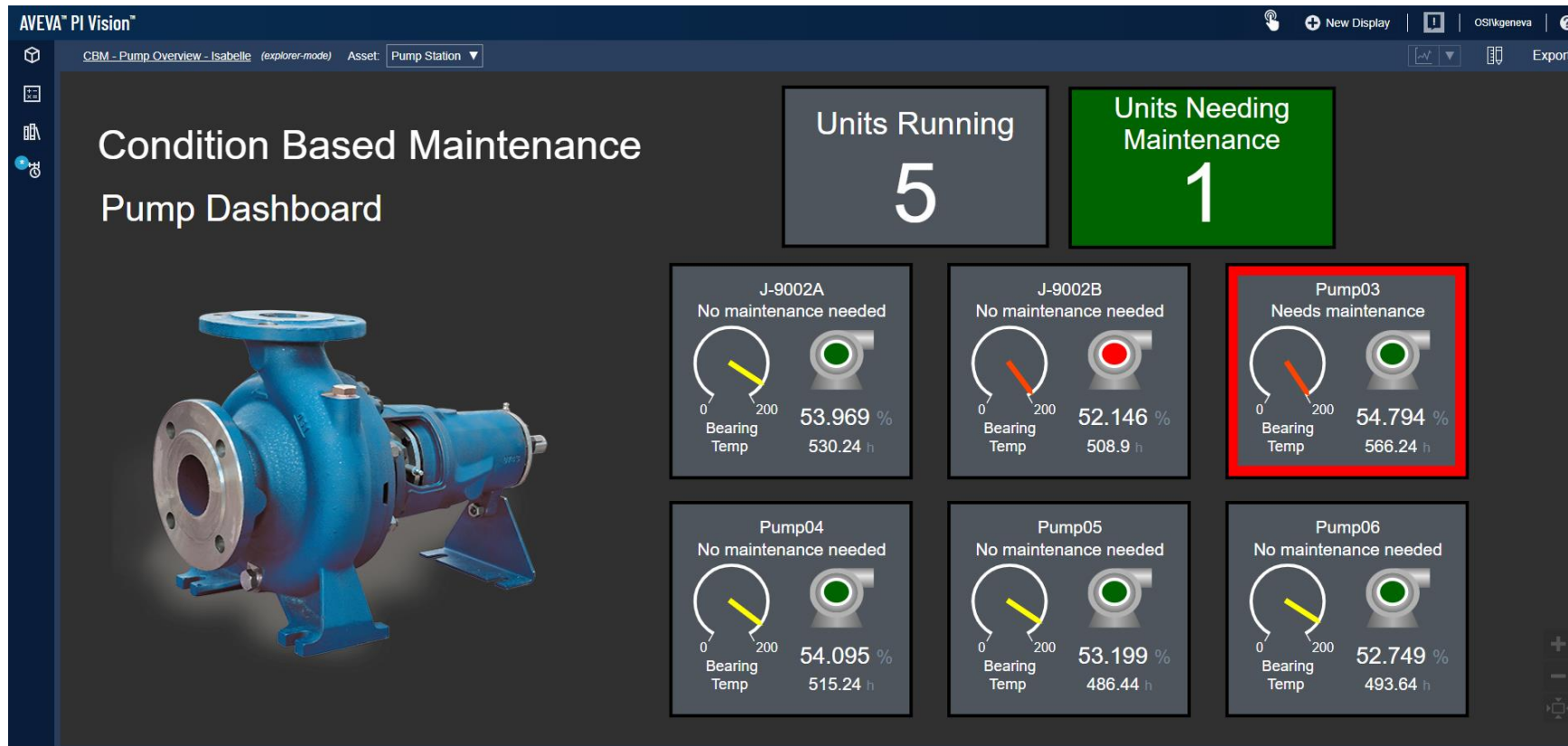
4 = Critical

2022-10-12 11:55:02 AM

1d

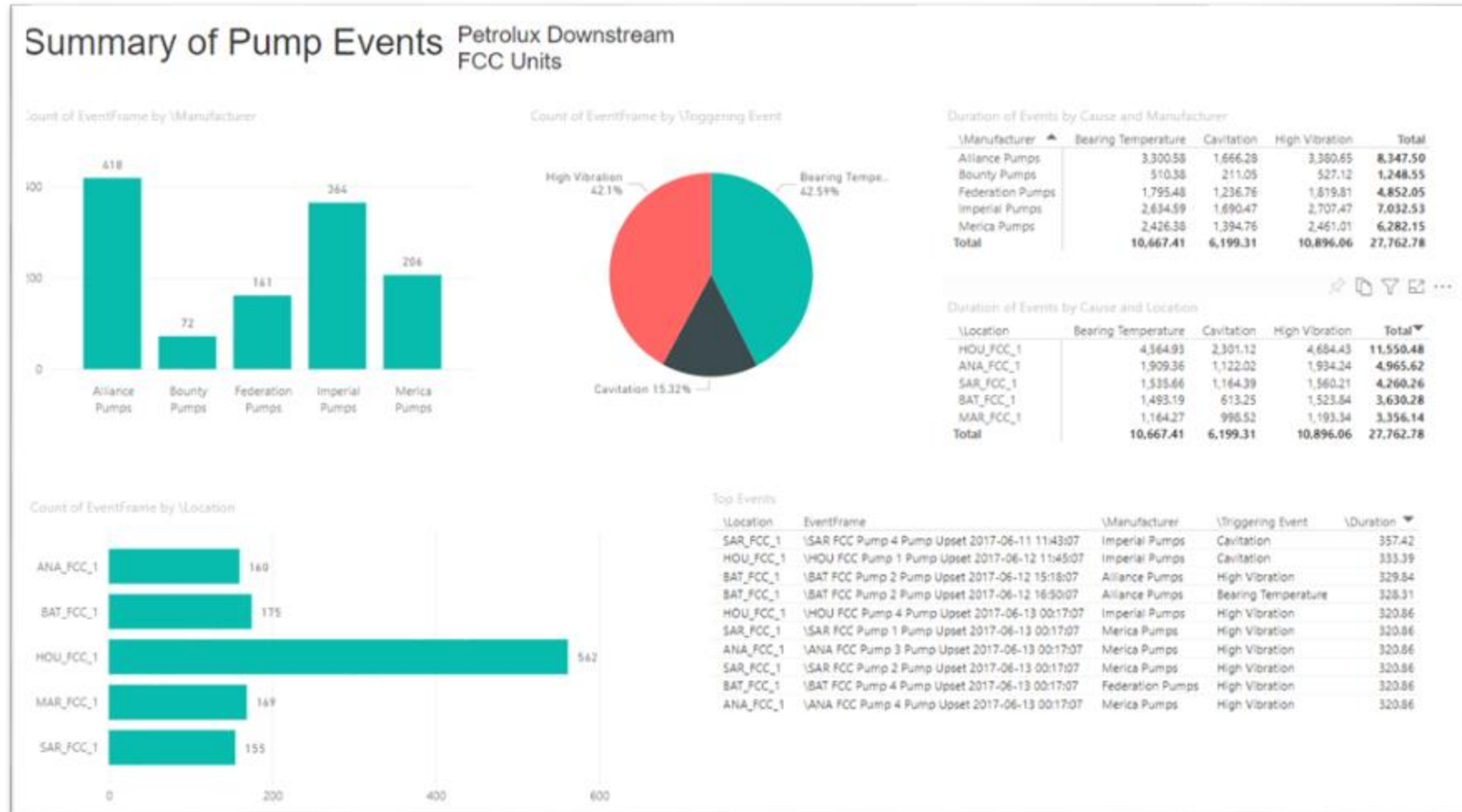
Alert users and close the loop

Notify and alert



Capture poor performance with event frames

Sum and analyze in PI DataLink or PowerBI



Overall Benefits of Condition Based Maintenance

- Reduced Downtime
- Reduction / elimination of Unplanned Failures
- Decreased Maintenance Costs
- Increased Asset Life
- Reduction in Collateral Asset Damages
- Situational Awareness and the ability to Save Precious Time
- Better Understanding of Asset Health
- Prioritization and Planning of Work Orders

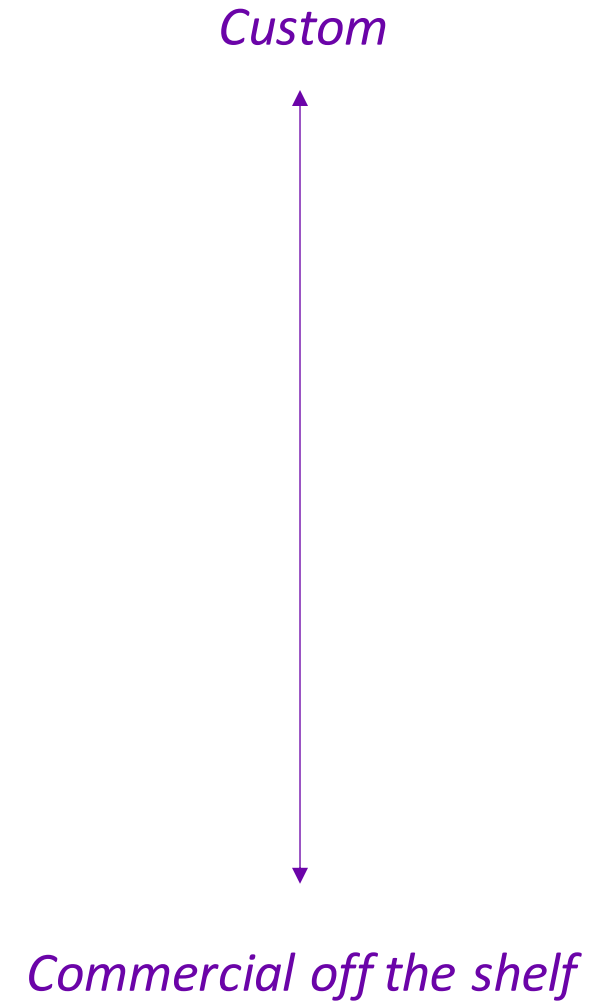


Artificial Intelligence - AI

AI – Predictive maintenance

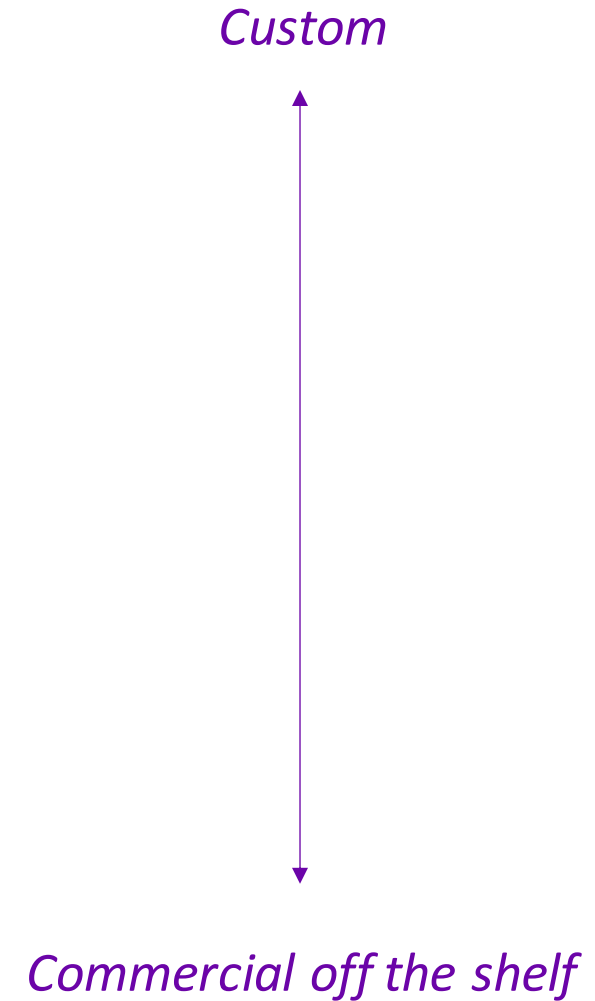
Starts with your people

- Full stack programmers and data scientists
- Only a data scientist
- No data science or programming skills



AI – Predictive maintenance

- PI System Access technologies
- PI Integrator for Business Analytics
- AVEVA Predictive Analytics
- AVEVA Data Hub and data sharing to partners



PI Web API

Secure RESTful access to PI Server data over HTTPS

- Send your data to Jupyter Notebooks
- Analyzed temperature control startup data to determine optimal start times for heating and cooling assets at the San Leandro Tech Center
- Findings showed theoretical savings of 5000 run hours per year for these assets

[Full presentation here](#)



AVEVA PI Integrator for Business Analytics sets foundation to accelerate advanced analytics and business intelligence

Data Sources

IIoT & sensors



Edge data stores



Plant control systems



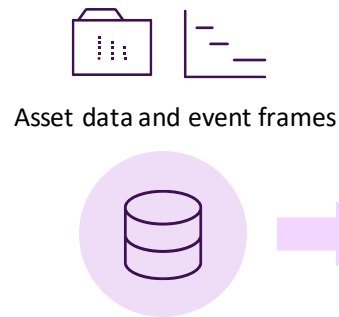
Historian



Remote assets



AVEVA PI Server

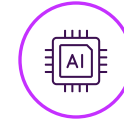


Transmit
Shape
Augment
Cleanse

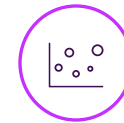
AVEVA PI Integrator for Business Analytics



AI/ML
platforms



3rd party
analytic tools



Enterprise
reporting

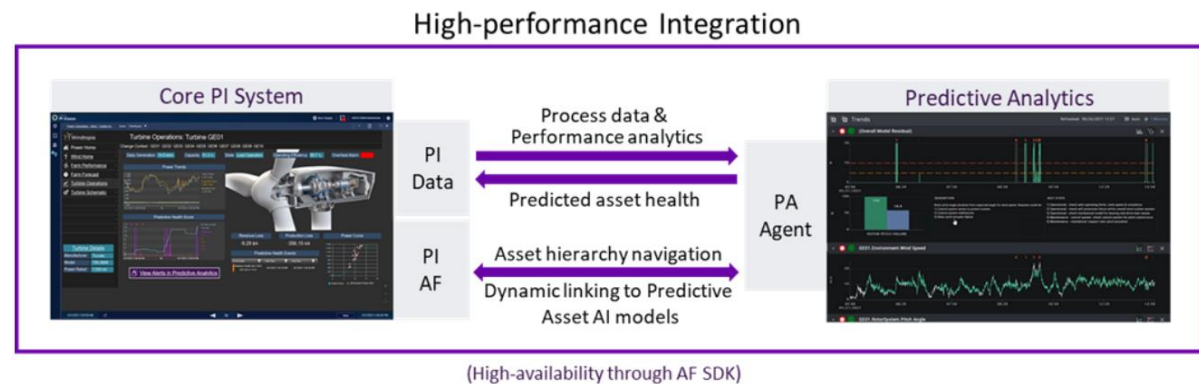


Apache Kafka
Amazon Kinesis
Amazon Redshift
Amazon S3
Cognos
Google Cloud Pub/Sub
Google Cloud Storage
Google Big Query
Hadoop
Hive
Microsoft SQL
MS Azure
Oracle
PowerBI
SAP HANA
Spotfire
SAS
Tableau

AVEVA Predictive Analytics

Prevent Equipment Failures

- Predictive monitoring of industrial equipment assets
 - Compressors, Pumps, Gearbox's, Motors, Turbines, Generators, etc.
- Achieve fast time to value through out-of-the-box predictive monitoring of industrial assets
- Developing unified PI + AI integration



EDF

North America

EDF runs fleet-wide monitoring of solar, wind and energy storage using AVEVA Predictive Analytics combined with PI System operational data management. The solution saved £1.5 million in a single early-warning catch.

Goals

- Combine the PI System™ operations data with AVEVA™ Predictive Analytics to mitigate major asset failures and optimize call-out decisions for maintenance crews.

Challenges

- With 39.1 million customers globally, EDF Group's brand image is closely linked to continuity of service
- Major failures could have been avoided before becoming severe, but the company did not have the capability to detect the early warning signs of future asset failures
- They resolved to create five monitoring centers in order to monitor over 300 fossil, nuclear and hydro plants

Results

- Avoidance of downtime resulting from equipment failure
- £1.5 million saved from a single early warning catch



Industry: Power generation

Solution: AVEVA™ PI System and AVEVA™, Predictive Analytics (formerly PRiSM)

“The PI System is designed to support our goals of operational intelligence. The idea is you build systems that take raw data and turn it into actionable information so you can make smarter decisions”

David Rodriguez Sr., Analytics and Intelligence Engineer, EDF Renewables



Duke Energy and Predictive Analytics

46 PI Servers and 11,000 models leveraging over 500,000 data points

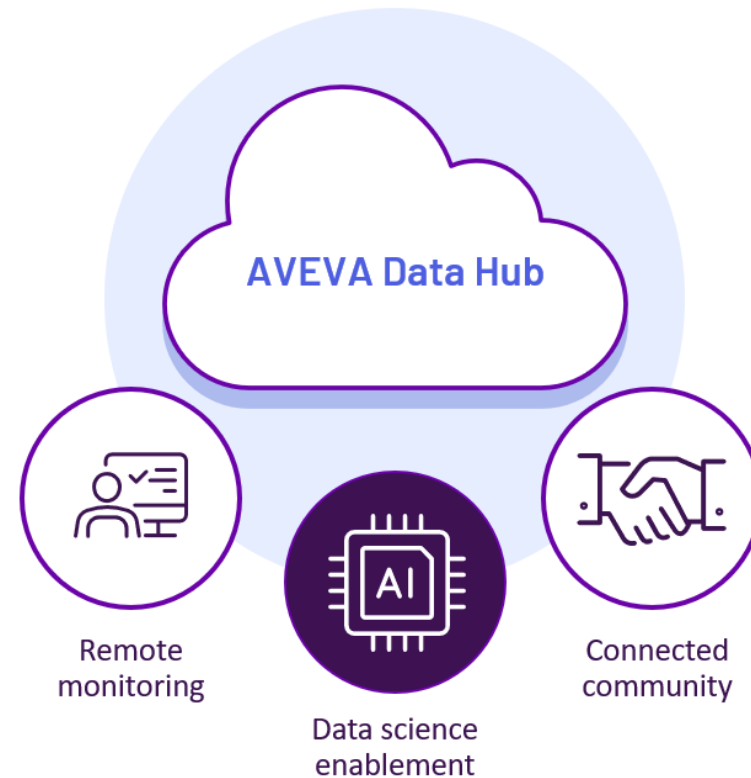
- Empower people with early warning notification of equipment problems
- Optimize assets with low-cost sensors and connectivity for high-fidelity data access enabling predictive maintenance
- Improve operations with contextualized insights
- Savings of over \$34 million in a single early catch event in 2016.
- AVEVA helps Duke Energy centrally monitor power generation assets with predictive asset analytics software

[Learn more](#)

AVEVA Data Hub and data sharing

Secure information sharing

- Selectively share operations data with trusted stakeholders
 - Key suppliers, equipment providers, data scientists, enterprise business analysts, and other authorized partners
- Connect external partners without giving network access



Renewable Energy Group (REG) and AVEVA Data Hub

Real-time external analysis on its centrifuge units

- Biodiesel equipment can have many recurring faults, and maintenance costs can also be high
- If this led to unplanned downtime, it could halt production, costing valuable revenue in addition to repairs
- Enabled simple, secure sharing of AVEVA PI System data using AVEVA Data Hub
- Solution can potentially reduce centrifuge downtime by up to 90%

[Learn more](#)



Putting data science predictors back into AVEVA PI Server

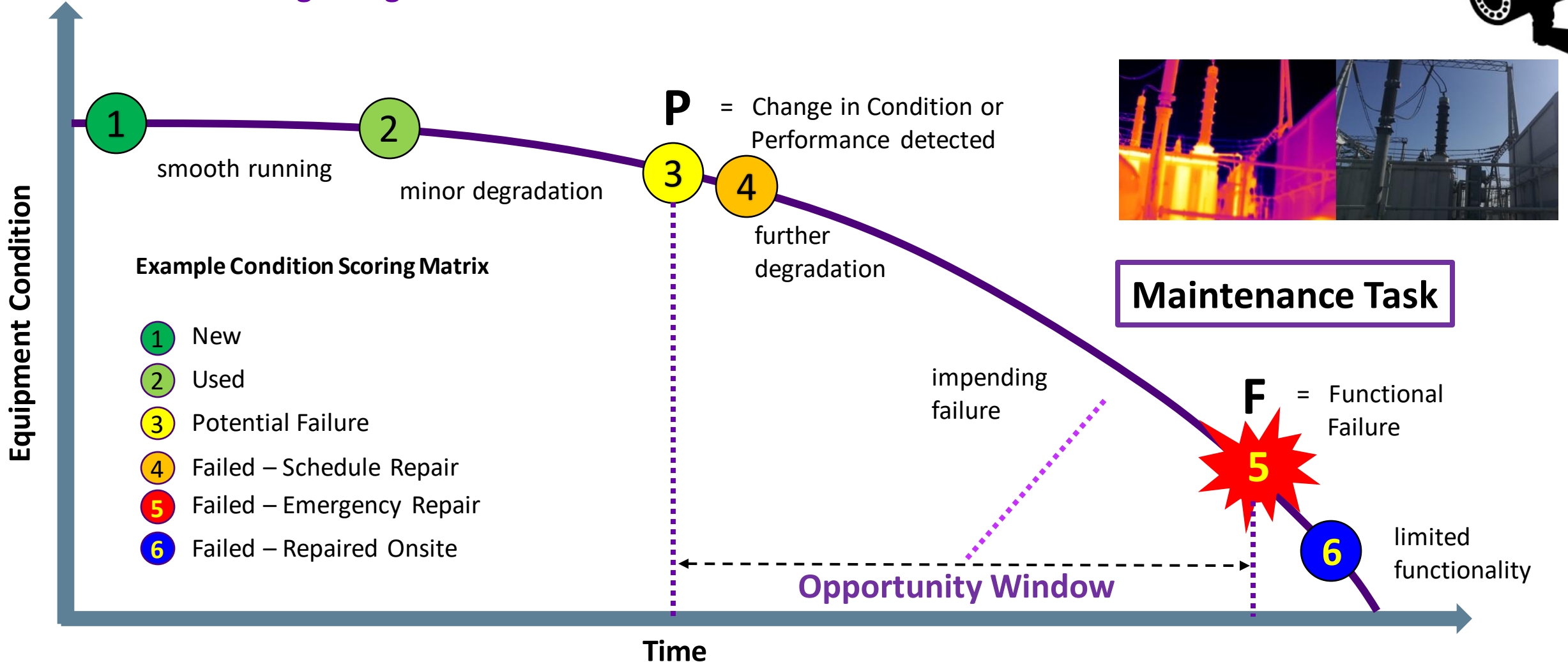
- Some predictive models have linear equation outputs
- Use Asset Analytics to create these calculations in PI, writing to future tags
- Notify on discrepancies between real time and predictive data



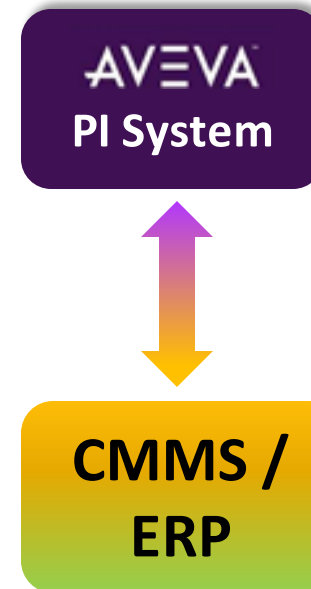
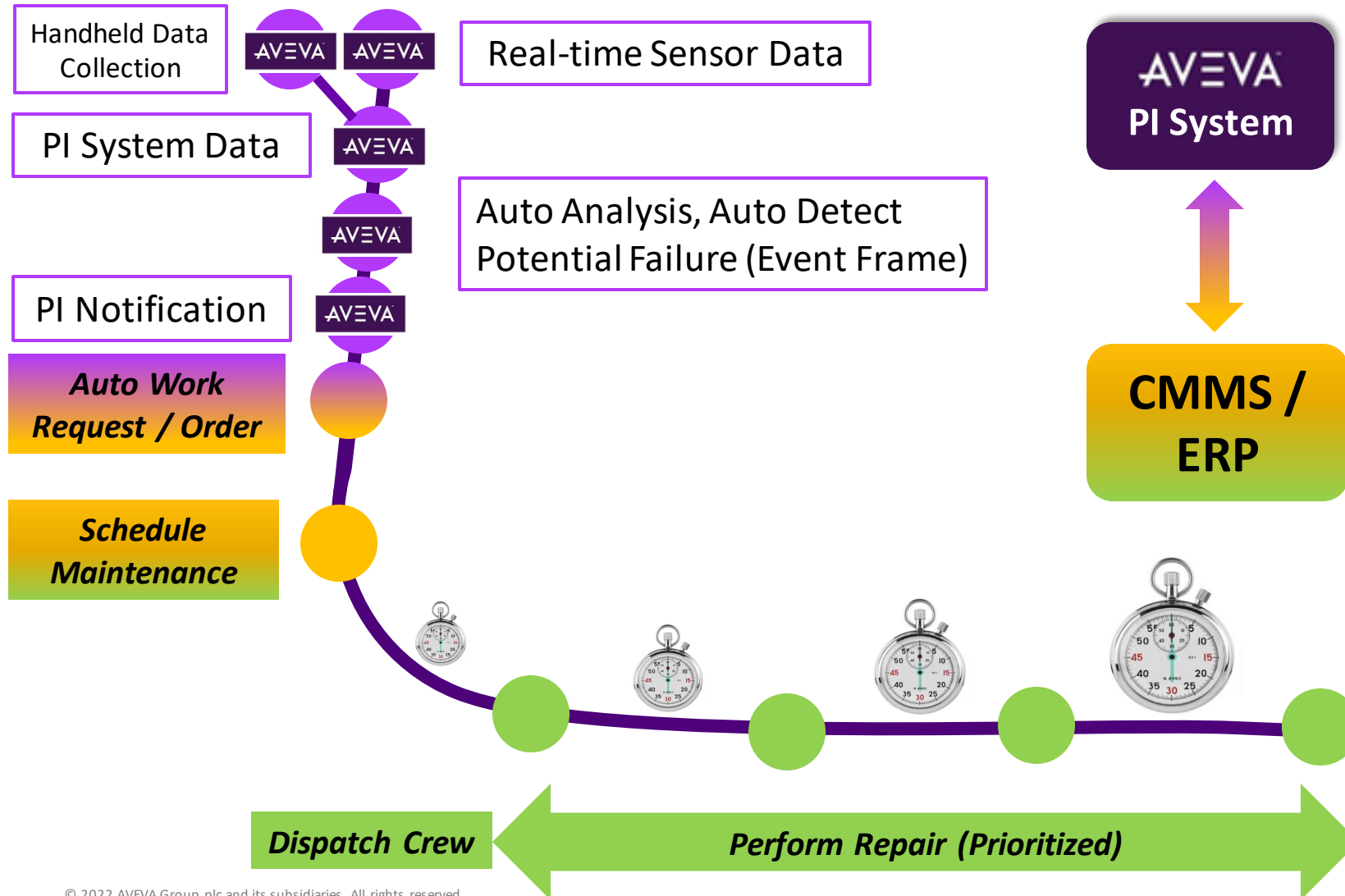
PI Data Egress to CMMS

Condition Monitoring / Incipient Failure Detection

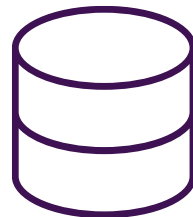
Condition Scoring Using Potential Failure “PF” Curve



Solution: Minimize Gaps Between Measurement and Action



Typical Timeline:
Seconds to Minutes
before made aware
of issue



PI Server

Middleware

CMMS

AVEVA

Real Example of CBM with PI System at Work

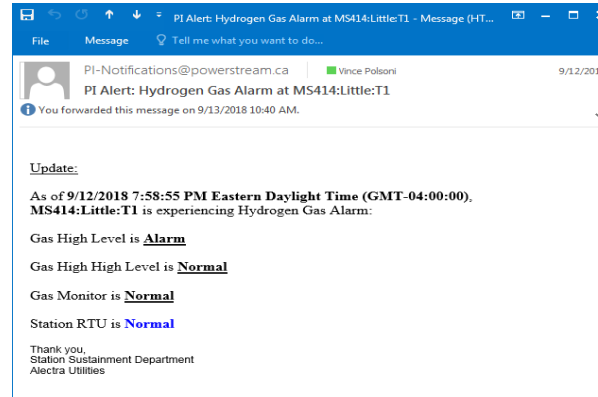
High Hydrogen Alarm on a 5 MVA Substation Transformer



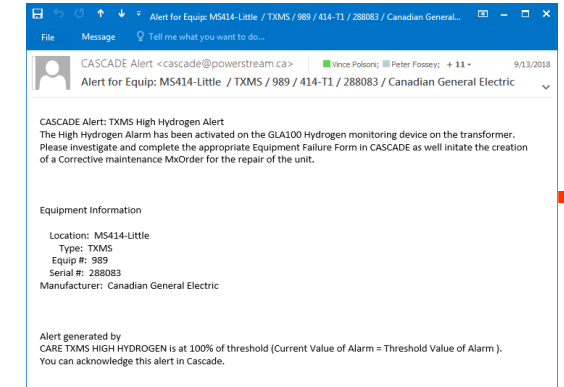
Hydrogen Sensor



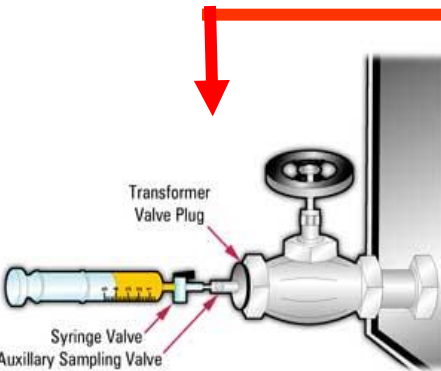
PI Vision Report



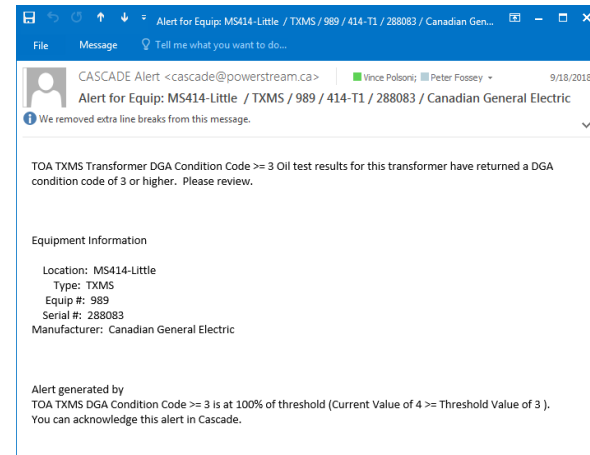
PI Notification – High Hydrogen



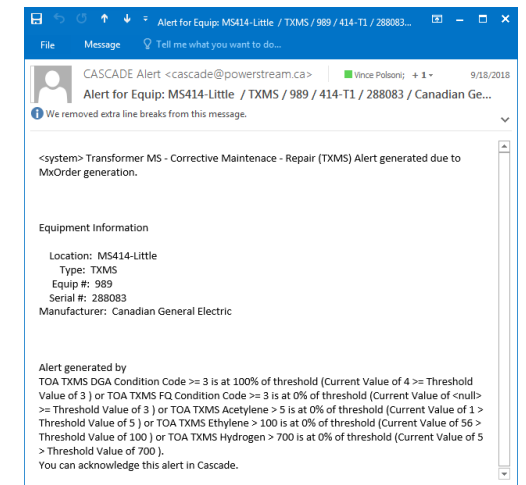
CMMS Alert – High Hydrogen



PI Vision Report



CMMS Alert – Poor Oil Condition



CMMS Auto – Repair Work Order

- Oil sample taken and analyzed by Laboratory
- Lab results entered into TOA4 (Advanced Analytics tool) for analysis
- Auto synch to CMMS and PI System

AVEVA



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Questions?

Please wait for the microphone
State your name and company



Please remember to...


Navigate to this session in the mobile
app to complete the survey.



Thank you!

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ABOUT AVEVA

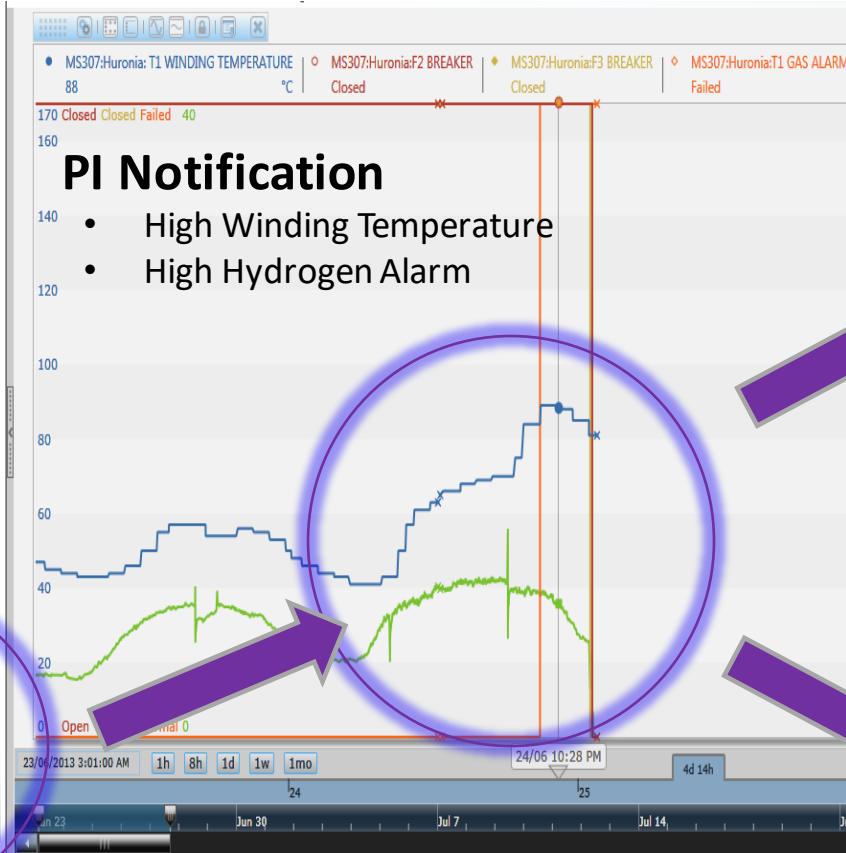
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Save - 10MVA 44kV-13.8 kV Transformer

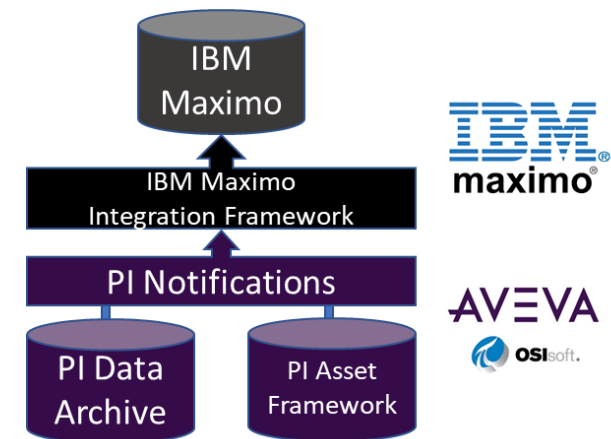
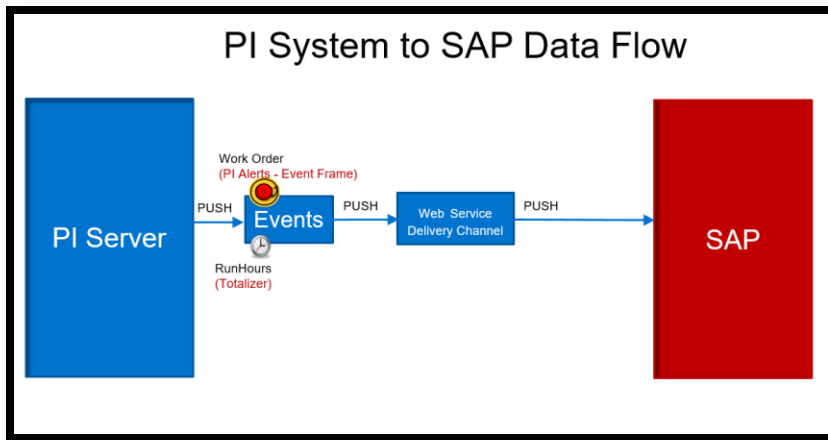


Cost Avoidance: \$500,000 averted

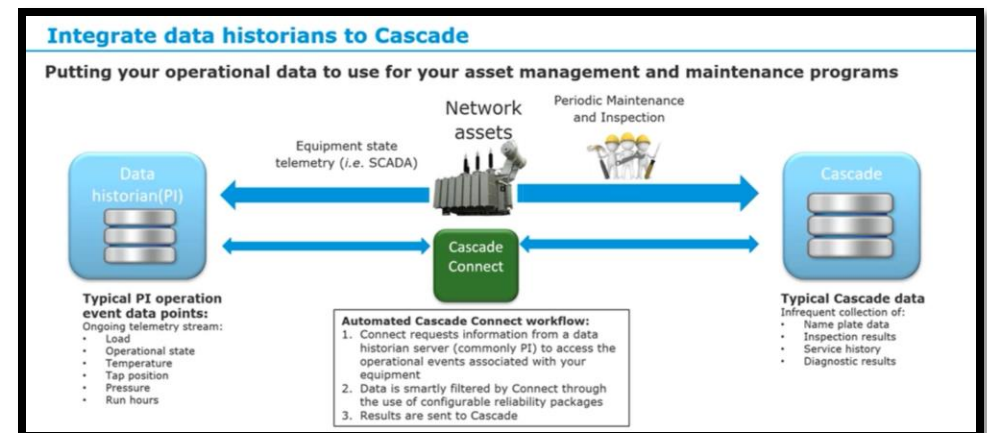
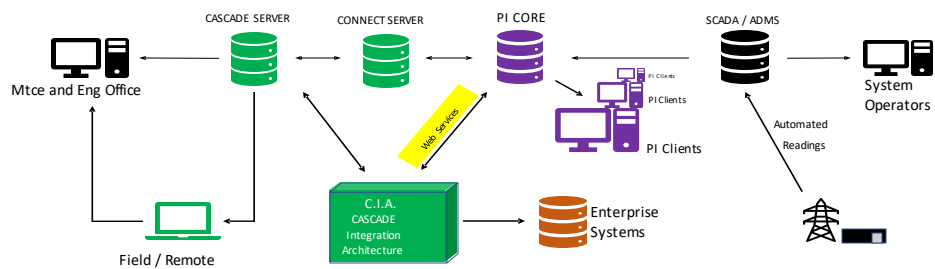
- Repair Cost: \$130,000
- No customer outages
- Transformer taken out of service before total failure, replaced with spare and repaired

Installed a Hydrogen Gas Monitoring Unit and connected to SCADA (PI and CMMS)





CASCADE CMMS Integration Methods



PI System to CMMS Data Flow

