

The background of the image is a dark blue gradient with a faint, stylized cityscape of San Francisco, including the Golden Gate Bridge and the Transamerica Pyramid. The OSIsoft logo is positioned at the top center.

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USERS CONFERENCE 2016

April 4-8, 2016 | San Francisco

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Delivering Business Value in Downstream O&G with Predictive Analytics and Machine Learning with PI AF and Microsoft Azure Machine Learning

Presented by **Tibor Komróczki**



Agenda – Delivering Business Value from the PI System

1. MOL Overview

2. Our Journey with the PI System and PI AF

3. Next Downstream Program

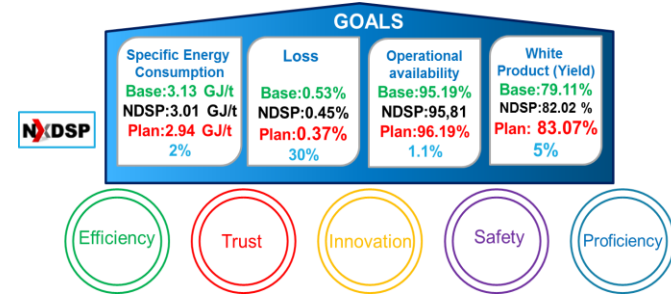
4. Integrity Operating Windows (IOWs)

5. Using PI AF for Forecasting & Predictive Analytics

6. Machine Learning with Microsoft Azure & the PI AF Foundation

7. Summary

8. Q&A





MOL Overview

MOL Group around the world

Operations in over 40 countries and employs almost 29,000 people worldwide

Production activities in 8 countries and exploration assets in 12 countries

Cultural alignment challenges - 10 time zones, 5 languages, and 3 units of measure

AFRICA

- Angola
- Cameroon
- Egypt



■ Countries of operation



EUROPE

- Austria
- Bosnia & Herzegovina
- Croatia
- The Czech Republic
- France
- Germany
- Hungary
- Italy
- Montenegro
- Poland
- Romania
- Russia
- Serbia
- Slovakia
- Slovenia
- Ukraine
- United Kingdom

ASIA

- Kazakhstan
- Kurdistan Region of Iraq
- Oman
- Pakistan
- Syria


MOL Downstream



- MOL operates 4 refineries and 2 petrochemicals plants, under integrated supply chain management.
- MOL also own a network of more than 1,900 service stations under 8 brands across 13 countries in Central & South Eastern Europe.

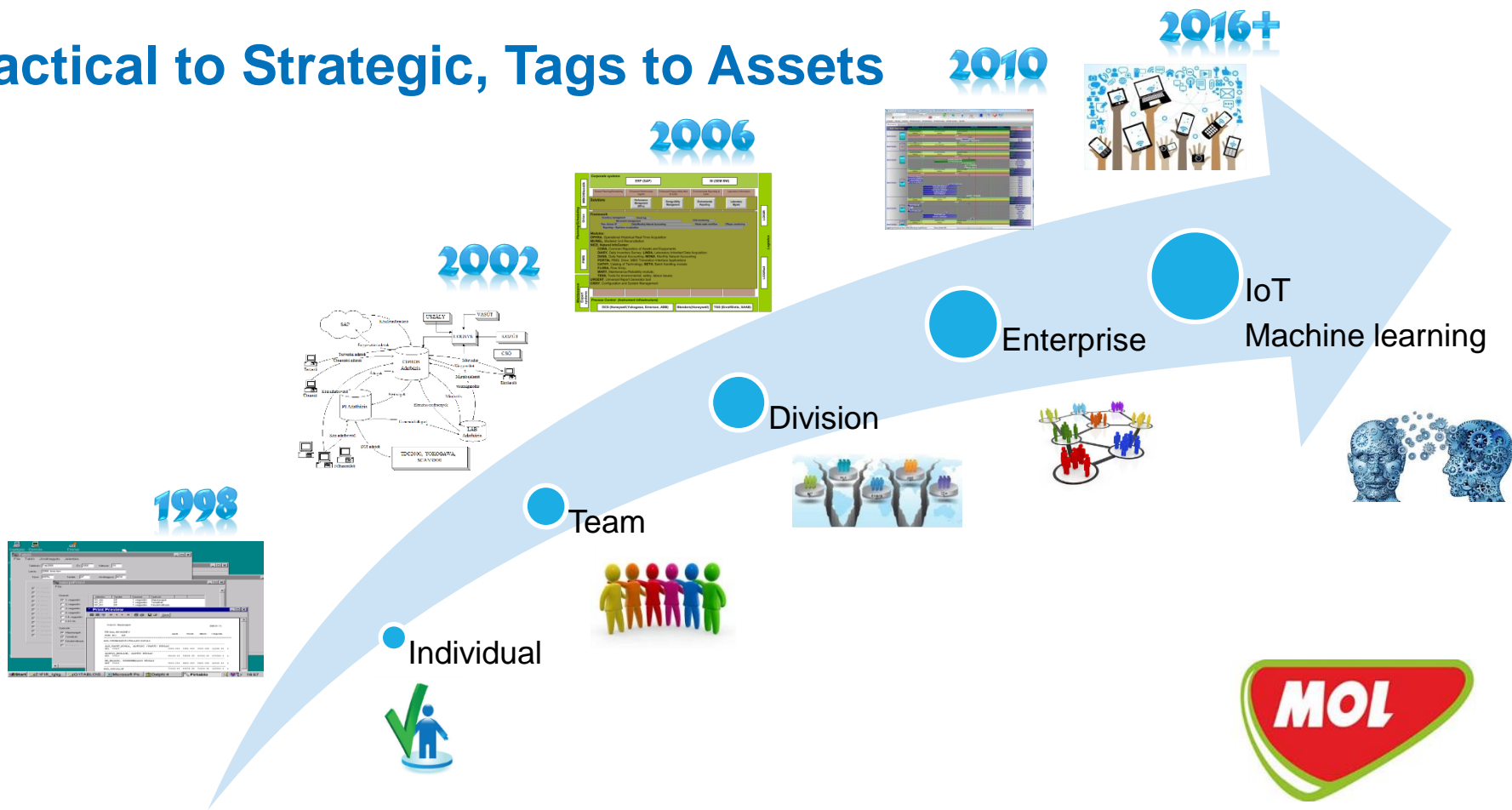


Element Categories	133
Element Templates	301
Elements	20914
Enumeration Sets	45
Event Frames	61012
Identities	95
Mappings	96
Notification Contact Template	154
Notification Templates	31
Notifications	5908

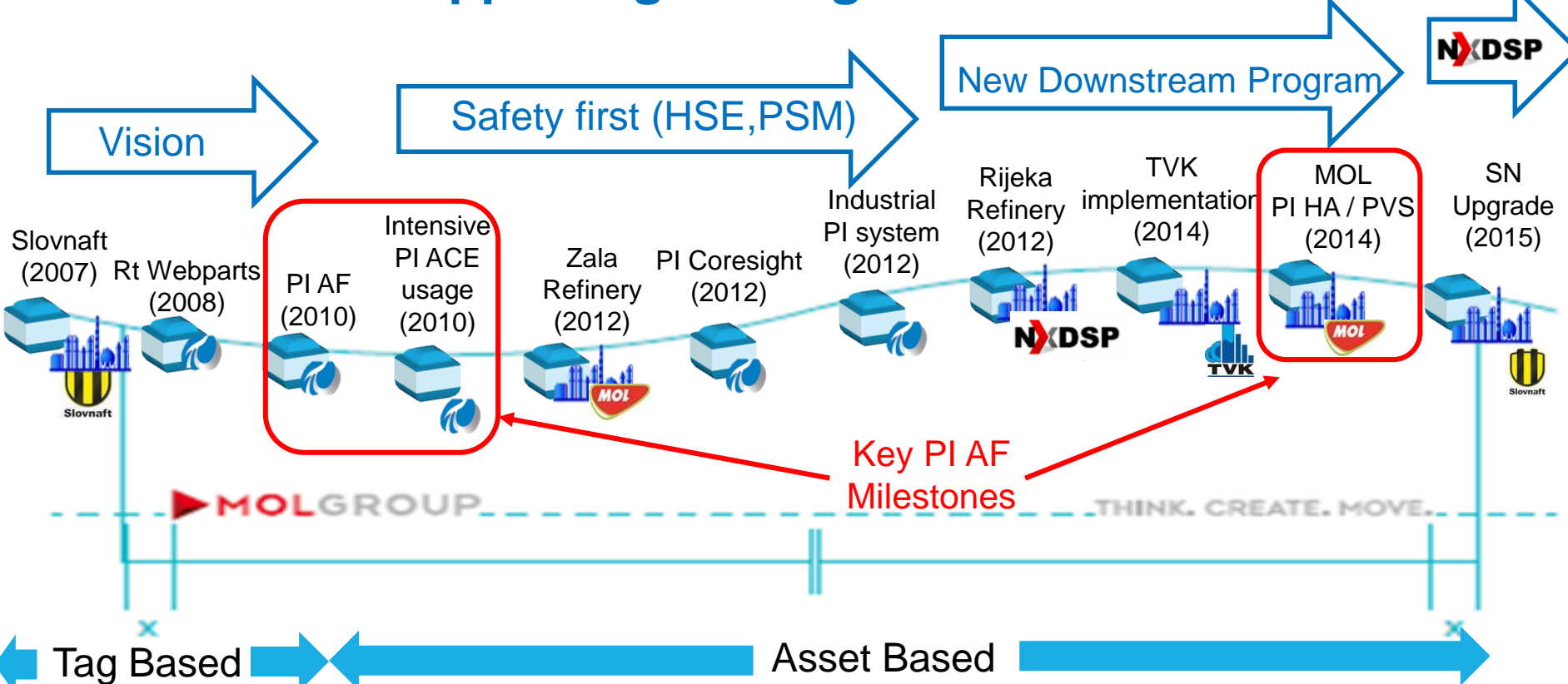


The MOL Journey with the PI System & PI AF

Tactical to Strategic, Tags to Assets

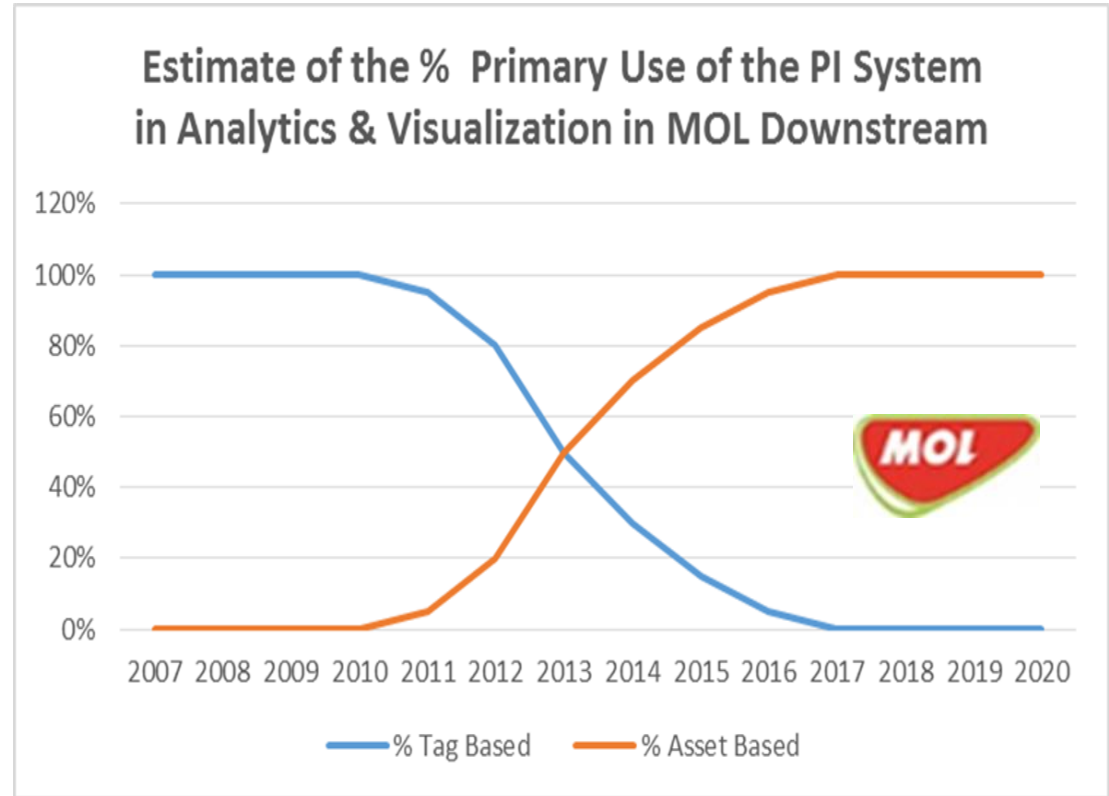


MOL Downstream / History II- Tactical Historian to Strategic Infrastructure- Supporting Strategic Business Initiatives

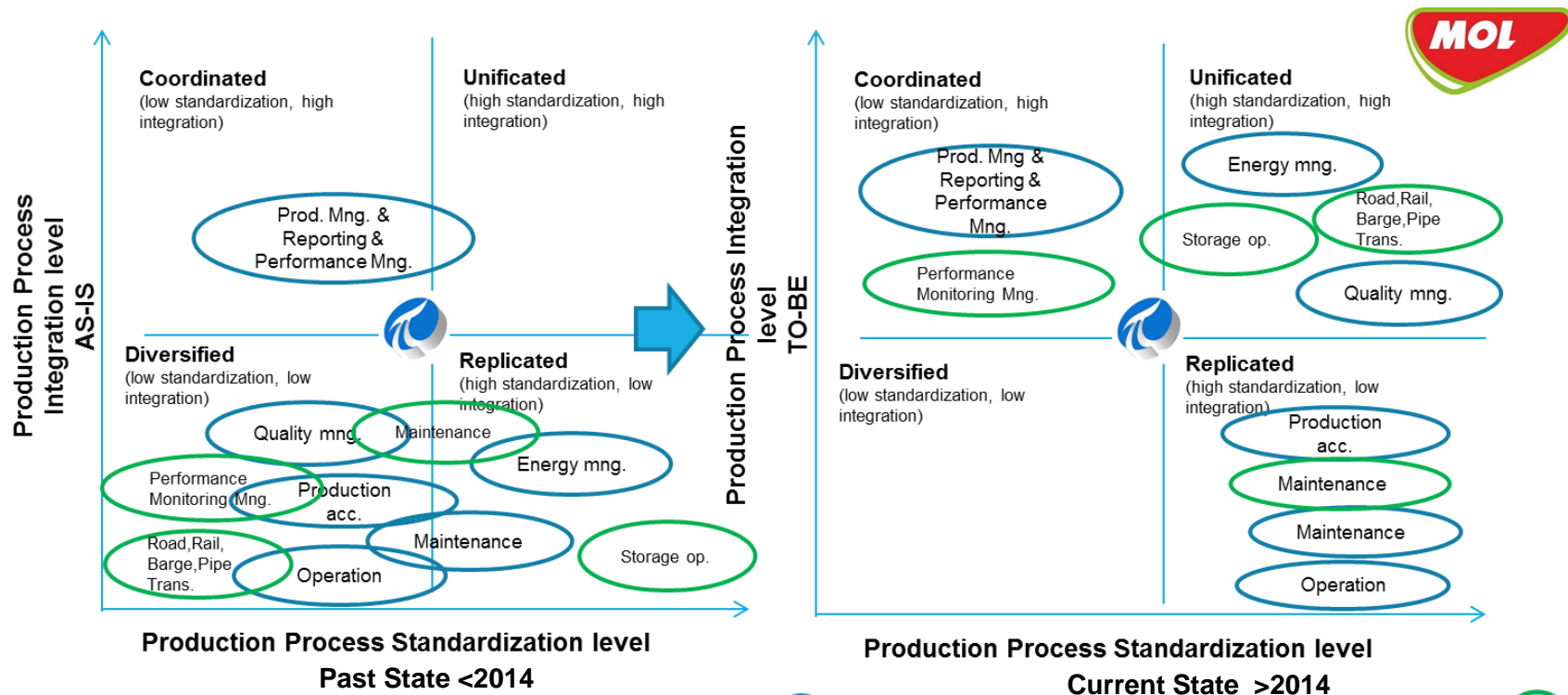


Moving to PI AF is a Vision and a Journey....."Just do It!"

- **Start PI AF with a key business initiative** and build, capability, momentum and awareness;
- **Market PI AF vision**, capability, and value to leadership with **alignment to strategic initiatives**;
- Leverage PI AF Jumpstarts, OSIsoft consultation, PI AF templates, SIs, etc. to **lower the barriers to PI AF use**;
- **Once PI AF critical mass** is achieved, adoption will accelerate and be seen as **transformative and strategic**;
- **MOL sees PI AF as enabling** to many PI System capabilities like PI Coresight, PI Connectors, Integrators, etc.;



Leverage of the PI System as an Integration & Applications Infrastructure to Simplify & Standardize the Applications & Solutions Portfolio



Production (Refinery & Petchem) & Logistic standardization

Safety and Asset Integrity

- Interlock statuses/DCS role tracking
- Operating envelopes
- Integrity Operating Windows (IOWs)
 - Corrosion control (HTHA)
- Alarm management
- Control rooms' temperature

Energy

- Energy Monitoring and Management
- Energy KPI breakdown (6 tiers)
- Column Energy Efficiency Dashboards
- Hydrogen, Utilities, and Energy balances
- Flaring

Yields

- Crude Blending Control
- Yield Optimization/Reporting
- Product Quality
- Analyser Reliability

Operational Optimization

- Plan vs Actual Analytics with Future Data
- NG and Fuel Demand gas forecasting
- Peak Electrical forecasting
- Normal mode of control loops
- APC monitoring
- PI AF and Sigmafine (PI AF) used for yield accounting & Material Movement



MOL Next Downstream Program

Next Downstream Program Goals

- Optimize maintenance costs and increase operational availability of MOL Group
- Move up one quartile energy intensity
- Black to white, increase of white product yield
- White yield to improve by 2.5%
- Flexibility between fuel and petchem products
- Diesel/Mogas from 2.4x to 2.76x



New DS program delivered USD 500mn Clean EBITDA improvement in 2014 vs 2011

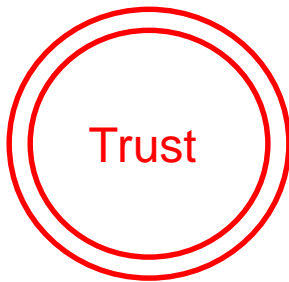
NxDSP targets **ADDITIONAL \$500M** increase by end of year 2016 vs 2014


Focus areas of New Downstream Program (NDP) and Next Downstream Program



NDS is targeting \$500M-550M EBITDA improvement by end of year 2016

GOALS







Supporting Refining Safely Excellence with Integrity Operating Windows(IOW)

Supporting Strategic Business Initiatives – Refinery Safety Excellence

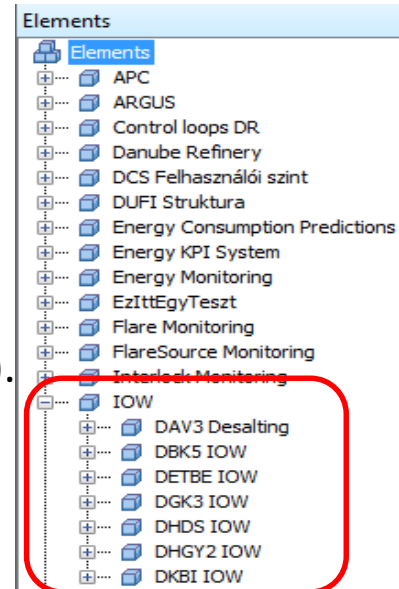


- On-line analyzers
- Interlock
- Flare
- Integrity Operating Windows (IOW)
- Operation Envelopes
- Proactive/Predictive analytics with PI AF
- Reports – PI DataLink
- Displays  **PI Coresight**
- Notifications  Notifications
- Safe operation
- Reduced shut-downs
- Availability/Utilization
- Integrity
- Loss of containment

Integrity Operating Window- Using PI Asset Analytics

IOW - Established limits for process variables (parameters) that can affect the integrity of the equipment if the process operation deviates from the established limits for a predetermined length of time – All in PI AF

- Immediate and comprehensive reliability improvement program for key units
- Perform Failure Mode and Effect Analysis (FMEA)
- Improve investigation & upgrade corrosion control
- Define Integrity Operating Windows of Critical equipment
- Prequalification of suppliers (cover more services, review the process).
- KPIs Review & Analysis of Cycle and Duration
- Reliability awareness improvement program and people competency



IOWs - Examples

Requires attention within specified timeframe	Requires drastic and/or immediate action
Standard	Critical
Heater Tube Skin Temperature	Boiler Feed Water Level
Crude Fractionator Dew Point Temperature	Hydro-process Reactor Temperature
pH of Crude Tower Overhead	Heater Tube Skin Temperature
Desalter Outlet Salt Content	Sulfuric Acid Strength in Alkylation

Proactive & Predictive Rather than Reactive Maintenance Leading to Improved Asset Integrity & Utilization



COMPETENCY DEVELOPMENT

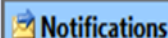
- Expert knowledge base
- Workshops
- Failure reports
- Laboratory background
- R&I

INSPECTION PROGRAM

- Advanced inspection techniques implementation
- Turnaround inspection plans preparation

CORROSION DATABASE

- Corrosion map
- Develop IOWs
Integrity Operating Window supported by PI client tools



MONITORING PROGRAM

- Comprehensive monitoring system development
- Operate monitoring system
- Expert analysis



MITIGATION PROGRAM

- Surface protection
- Protection by chemicals
- Cathodic protection
- Material changes

Danube Refinery Crude Distillation Unit IOW – AF Analytics

\\MOLSZHBFINAF01\RtCenter - PI System Explorer (Administrator)

File Search View Go Tools Help

Database Query Date Back Check In Refresh New Element Search Elements

Elements

- Flare Monitoring
- Integrity Operating Windows
 - DOESTB Block
 - DAV3 Unit
 - Ammoniatartalom 121 elfolyo vizében
 - Ammoniatartalom a 122 elfolyo vizében
 - Hőmérséklet 101 fején
 - Kloridtartalom 121 elfolyo vizében
 - Kloridtartalom 122 elfolyo vizében
 - pH 121 elfolyo vizében
 - pH 122 elfolyo vizében
 - Sótartalom 138 II. lépcsőből
 - DKBI Unit
 - DMOHAB Block
 - DREHIB Block
 - Interlock Monitoring
- Element Searches

DAV3 Unit

General Child Elements Attributes Ports Analyses Version

Name: IOW Unit Rollup

Description:

Categories: IOW General|IOW Unit

Analysis Type: ☐ Expression ☒ Rollup ☐ Event Frame Generation

Rollup attributes from

☒ Child elements of DAV3 Unit

☐ This element - DAV3 Unit

To select attributes set criteria below

Attribute Name: Exceedance State

Attribute Level: Root Level

Attribute Category:

Element Category:

Element Template:

Select the function(s) to write to an attribute Evaluate

Function	Output(s)	Value
<input checked="" type="checkbox"/> Sum	Count of Exceed	
<input type="checkbox"/> Average		
<input type="checkbox"/> Minimum		
<input type="checkbox"/> Maximum		
<input checked="" type="checkbox"/> Count	Count of Param	
<input type="checkbox"/> Median		
<input type="checkbox"/> Population standard deviation		
<input type="checkbox"/> Sample standard deviation		


Sample Child Element: Ammoniatartalom 121 elfolyo vizében

Group By: None

Name	Parent Element	Categories
✓ Exceedance State	Ammoniatartalom 121 elfolyo vizében	Results
Block ID	Ammoniatartalom 121 elfolyo vizében	General Attributes
Current	Ammoniatartalom 121 elfolyo vizében	Process Data
Desc	Ammoniatartalom 121 elfolyo vizében	General Attributes
Exceedance Time Ratio	Ammoniatartalom 121 elfolyo vizében	Time Cumulated Data
Functional Location	Ammoniatartalom 121 elfolyo vizében	General Attributes
HI Limit	Ammoniatartalom 121 elfolyo vizében	Limit
Limit Status	Ammoniatartalom 121 elfolyo vizében	Results
LO Limit	Ammoniatartalom 121 elfolyo vizében	Limit
Name	Ammoniatartalom 121 elfolyo vizében	General Attributes
Unit ID	Ammoniatartalom 121 elfolyo vizében	General Attributes
Unit Operating	Ammoniatartalom 121 elfolyo vizében	Configuration Parameters

Danube Refinery Crude Distillation Unit – IOW Visualization Augmentation of Opralog





Corrosion & Integrity Predicting High Temperature Hydrogen Attack(HTHA)

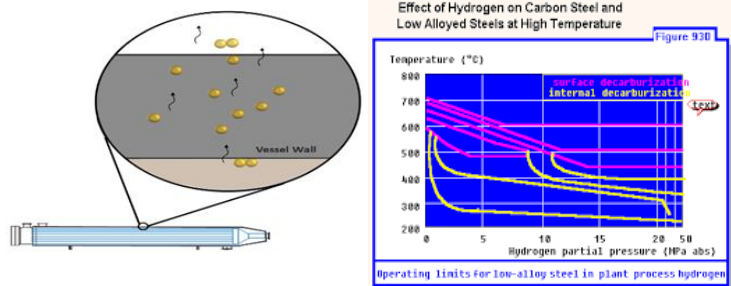
Improving Asset Integrity with Advanced Corrosion Analytics

Category: Auxiliary Attributes		
H2 Partial Pressure		3.98419110750565 bar
PA		510
PB		0
PC		0
PD		-183
PE		0
PF		0
PG		0
Pressure Abs		20.6218204498291 bar
TA		615
TB		-1.24
TC		0
TD		0
TE		0
TF		0

- High Temperature Hydrogen Attack (HTHA)
- $f \times$ (metallurgy, temperature, hydrogen partial pressure(PP), length of exposure)

Developed PI AF template that:

- Determine partial pressure
 - Attribute of pipe class
 - Temperature and length of exposure limits
 - Total time above Temp and PP
 - Alerts/notification/event frame
- Tested and rolled out in 6 units < 1 week
 - Expanded to all plants in 2015.



Nelson Temperature Limit LookUp Function and LookUp table

Table Lookup Data Reference

Table: IOW HTHA Nelson Curves T lim

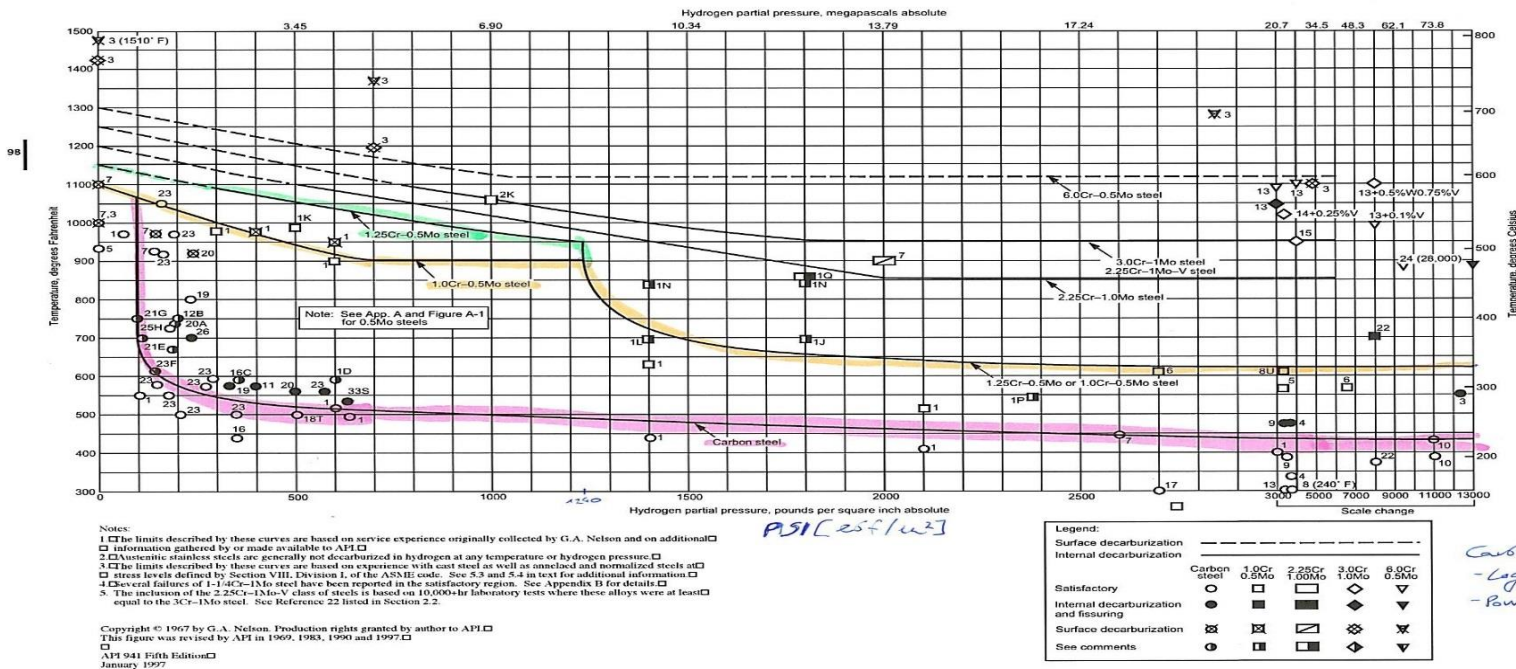
Material Type	TMIN	TMAX	PA	PB	PC	PD	PE	PF	PG
1,25Cr-0,5Mo	614.8208452	1000	0	0	0	0	0	1	0
1,25Cr-0,5Mo	508.7477369	614.8208452	495.1655322	-0.805381822	0	0	0	1	0
1,25Cr-0,5Mo	322.0725521	508.7477369	80	0	0	1	1021.226335	-320.6541822	1
1,25Cr-0,5Mo	-100	322.0725521	800	0	0	0	0	1	0
1Cr-0,5Mo	588.9601409	1000	0	0	0	0	0	1	0
1Cr-0,5Mo	482.2222222	588.9601409	260.3998906	-0.442134998	0	0	0	1	0
1Cr-0,5Mo	322.0725521	482.2222222	80	0	0	0	1021.226335	-320.6541822	0
1Cr-0,5Mo	-100	322.0725521	800	0	0	0	0	1	0
Carbon Steel	576.6666667	1000	6.89475729	0	0	0	0	1	0
Carbon Steel	354.4274913	576.6666667	6.89475729	0	0	0	0	1	0
Carbon Steel	197.1283418	354.4274913	6	0	-0.043155263	15.18420899	0	0	1
Carbon Steel	-100	197.1283418	800	0	0	0	0	1	0

Value to return when no matching row found: -1000

Value to return when NULL result found: -1000

OK Cancel

Use of Excel Solver to Determine Equation from a Data Set



$$N_{\text{emp}} = \frac{1}{x+B} + A$$

$$y = \frac{y_{\text{max}} \cdot x}{B+x} + A$$

$$y = y_0 \cdot e^{-Bx} + A$$

$$-(A - e^{-B})$$

$$A + e^{-B}$$

$$1 - e^{-Bx}$$

Carbon steel: $PH_2 - T(°C)$ equation

$$-Log: -29.54 \ln(X) + 462.23$$

$$B^2 = 0.252$$

$$-Power: y = 549.88x - 0.103$$

$$B^2 = 0.2313$$

$$y = A(1 - e^{-B(P-2.5)}) + 482$$

$$1 - e^{-B(P-2.5)}$$

$$0 - (1 - e^{-B(P-2.5)})$$

Equation 1

$$y = -200.6x + 615 \quad y_2 = -188M - e^{-1.047(PH_2 - 2.543)} + 510$$

$$y_1 = -0.155x + 583 \quad y_2 = 482 \quad y_3 = -150(1 - e^{-0.655(PH_2 - 2.543)}) + 482$$

$$y = -29.54 \ln(X) + 462.23$$

Illustrative PI AF IOW Predictive Analytics Configuration

PI System Explorer (Administrator)

File Search View Go Tools Help

Database Query Date Check In Refresh New Element

Elements

Elements

- Flare Monitoring
- Integrity Operating Windows
 - DDSTB Block
 - DMCHAB Block
 - DBK5 Unit
 - DET6 Unit
 - Hidrogénozo reaktor
 - DGK3 Unit
 - DHDS Unit
 - DREHB Block
 - DSR2 Unit
 - Interlock Monitoring
 - Element Searches

Hidrogénozo reaktor

General Child Elements Attributes Ports Analyses Version

Name: HI Limit

Description:

Categories: IOW General IOW HTHA Limit Calculation

Backfilling

HI Limit

IOW HTHA Exceedance State Calculation

Name Expression

ExceedanceStateCalculation IF (('Current'>'HI Limit Temperature') AND 'Unit Operating'=1) THEN 1 ELSE 0

Add a new expression

HI limit

IOW HTHA Exceedance state calculation

IOW HTHA

PI Coresight

Area	IOW parameter / PI tag	Act	<	HI Limit	s
DBK5 IOW	Kénmentesítő reaktor betáp előmelegítő cseppfogó - Hőmérséklet DBK5RT12017.DACA.PV	229,2	<	270	
DBK5 IOW	Kénmentesítő reaktor betáp előmelegítő cseppfogó - Nyomás DBK5RPI2038.AIA.OUT.VALUE	19,342	<	23	
DBK5 IOW	Kénmentesítő reaktor betápkilépő hőcserélő - Hőmérséklet DBK5RT12015.DACA.PV	156,8	<	270	
DBK5 IOW	Kénmentesítő reaktor betápkilépő hőcserélő - Nyomás DBK5RPI2037.AIA.OUT.VALUE	19,6589	<	23	
DBK5 IOW	Szelektív hidrogénozo reaktor betáp előmelegítő - Hőmérséklet DBK5RT11031.DACA.PV	149,9	<	199	
DETBE IOW	Hidrogénozo reaktor- Hőmérséklet DMT6TTH054.PV	89,8393	<	200	
DGK3 IOW	Cirkulációs gáz/Friss H2 dús gáz hőcserélő - osó - Hőmérséklet DGK3KT1592.PVA	142,542	<	210	
DGK3 IOW	Cirkulációs gáz/Friss H2 dús gáz hőcserélő - köpeny - Hőmérséklet DGK3KT1553.PVA	178,208	<	230	
DGK3 IOW	Cirkulációs gáz/könnnyő alapanyag hőcserélő - Hőmérséklet DGK3KT1555.PVA	136,64	<	200	
DHDS IOW	Cirk. Gáz előmelegítő 103/1- köpeny- Hőmérséklet DHD5HT1179.PV	223,311	<	240	
DHDS IOW	Cirk. Gáz előmelegítő 103/2- köpeny- Hőmérséklet DHD5HT1290.PV	231,833	<	240	



Natural Gas Purchase Optimization

Business Challenge – Exceeding NG Daily Consumption Peak

New MOL Hungary Enterprise contract

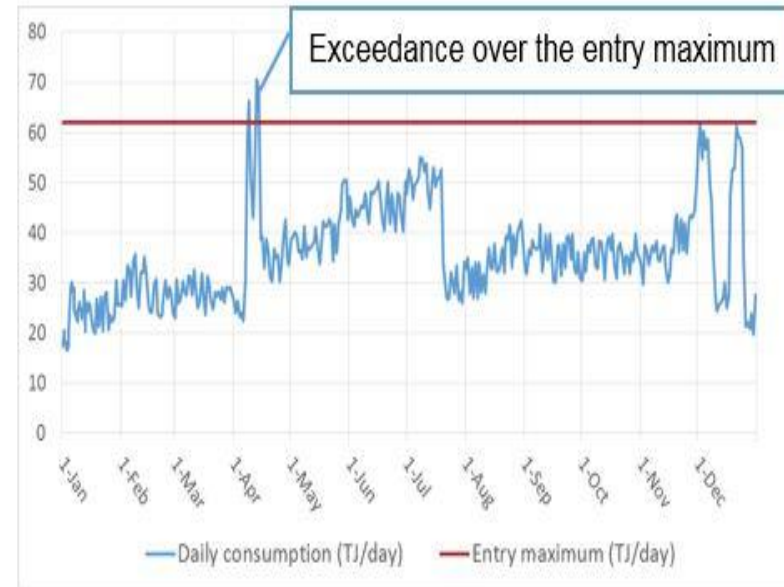
- Entry Maximum decrease → potential saving
- Entry Maximum fixed for two years
- High penalty on any exceedances
- Fluctuation in consumption

Objective

- Decrease the entry maximum
- Eliminate the chance of exceedance



- Eliminate the peaks of consumption




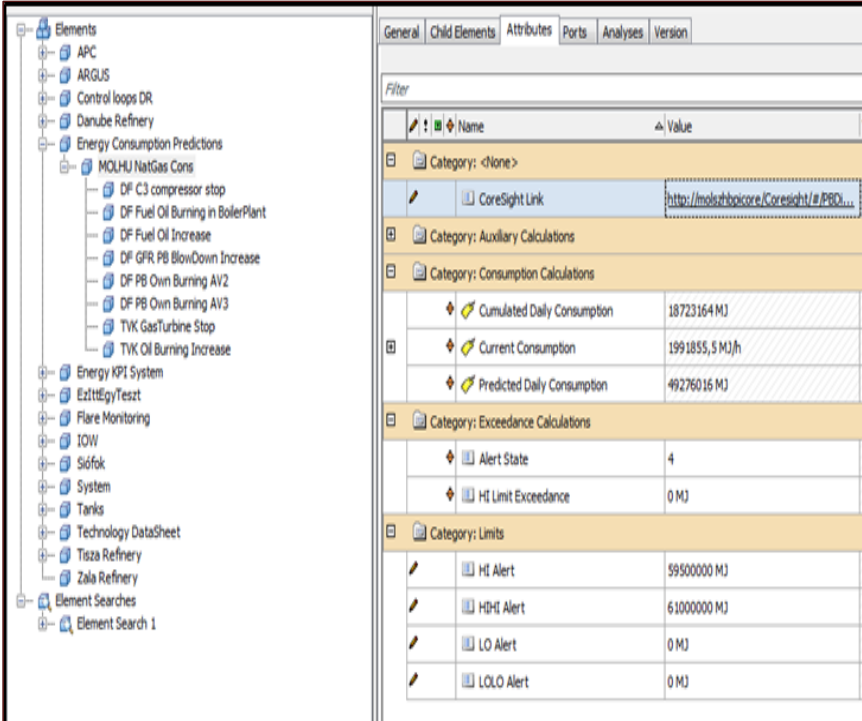
PI AF Predictive Analytics – Forecasting NG Supply/Demand

Daily & hourly forecasting system

- Continuous calculation
- Based on consumption data
- **PI Asset Analytics**

New tool to improve cooperation

- Different sites accross the country
- Alerting system
- Up-to-date, valid information
-  **PI CoreSight**



The screenshot displays the PI CoreSight interface. On the left, a tree view shows the hierarchy of elements, including APC, ARGUS, Control loops DR, Danube Refinery, Energy Consumption Predictions, MOLHU NatGas Cons, and various fuel oil and gas turbine components. On the right, a table with tabs (General, Child Elements, Attributes, Ports, Analyses, Version) shows data for selected elements. The table is filtered by 'Name' and displays values for various consumption and alert categories.

Name	Value
Category: <None>	
CoreSight Link	http://molshbpicore/CoreSight/#PRD...
Category: Auxiliary Calculations	
Category: Consumption Calculations	
Cumulated Daily Consumption	18723164 MJ
Current Consumption	1991855,5 MJ/h
Predicted Daily Consumption	49276016 MJ
Category: Exceedance Calculations	
Alert State	4
HI Limit Exceedance	0 MJ
Category: Limits	
HI Alert	59500000 MJ
HIHI Alert	61000000 MJ
LO Alert	0 MJ
LOLO Alert	0 MJ

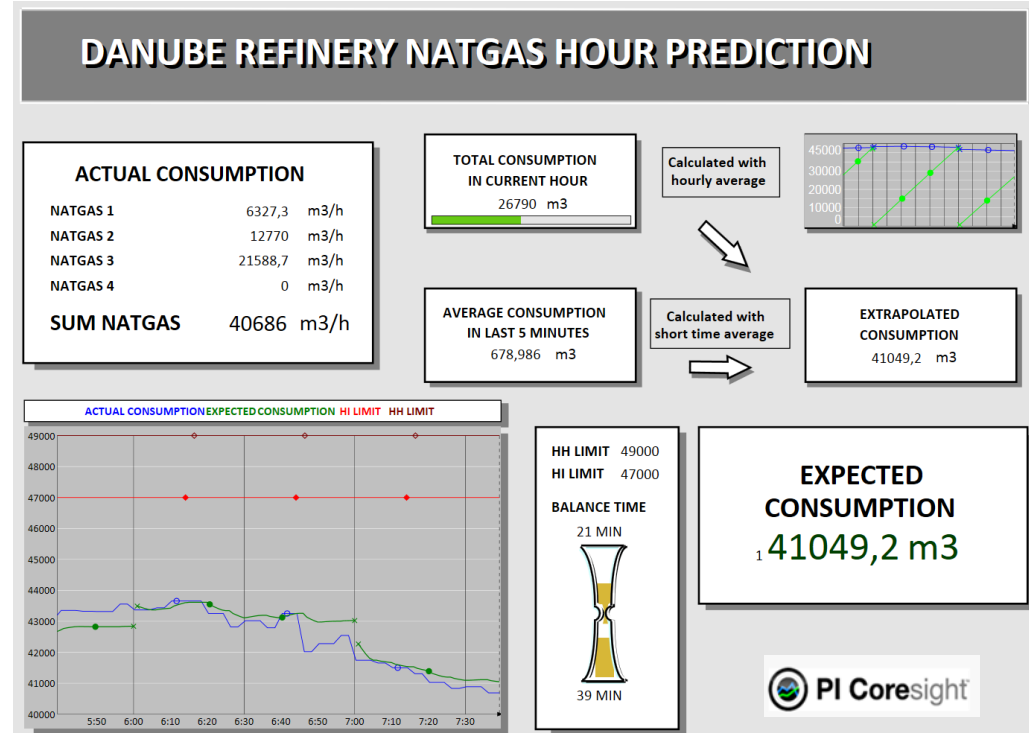
Building on Success – Daily to Hourly Natural Gas Prediction

The planned system

- Online calculation (extrapolation in every hour)
- Input:
 - online sensor data
 - static data and
 - periodically filled manual data

Tools

- PI Asset Analytics
- AF Tables and Lookup calculations avoidance of unnecessary alerts via smoothing of forecasts within the hour



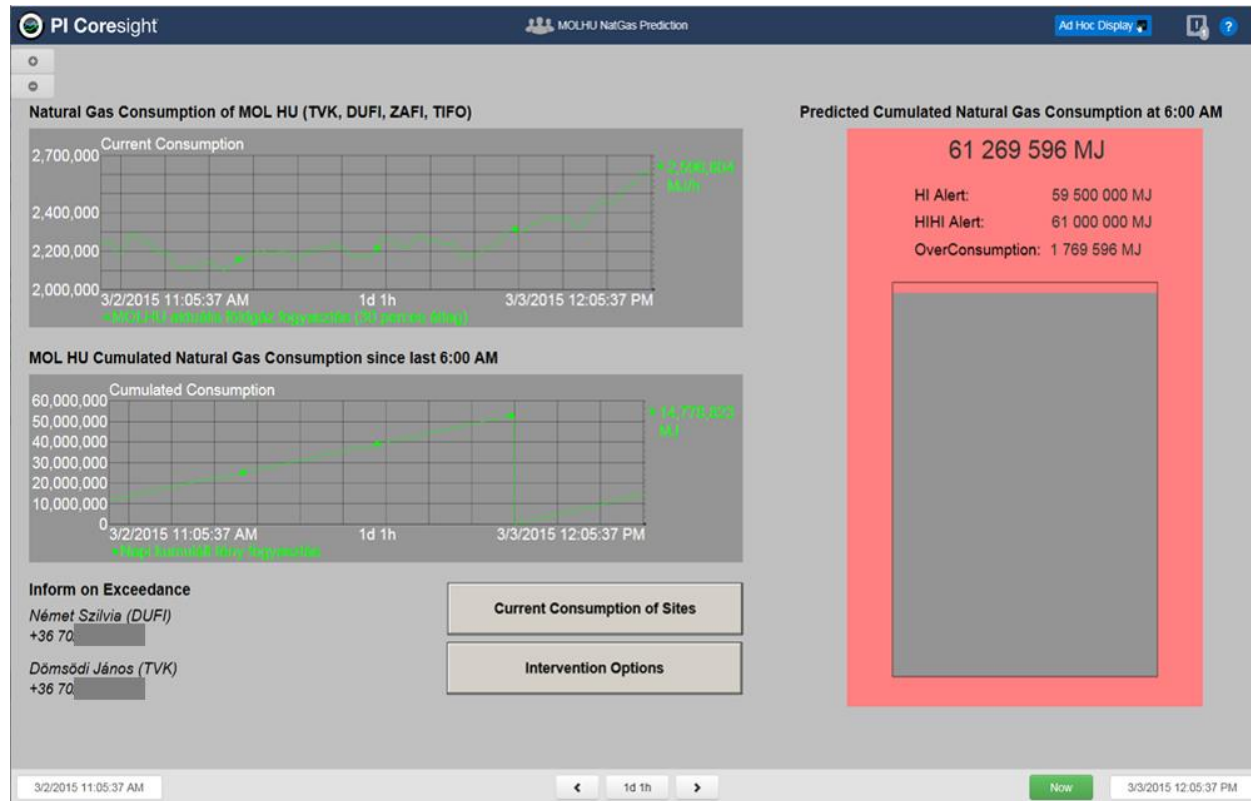
Results - The Working System

Alerting

- Conversation started immediately after the alert
- Decision has been made in around 30 minutes
- Employees were informed

Business Impact

- 2 TJ/day Entry
- Maximum decrease
- \$230k/yr savings



Continuing the PI AF Journey

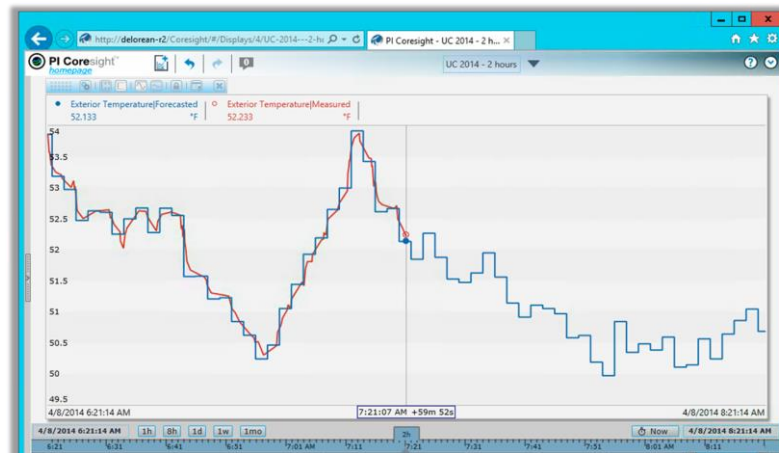
Calculation Accuracy


- The accuracy of the forecasting calculation can be increased (Tuning)
- Involve daily scheduling data or unit operation mode
- **Use of Future Data and PvA Analytics**



Extending to Other Energy Types

- Increase the planning accuracy in case of other energy types
- Installation is very simple and fast

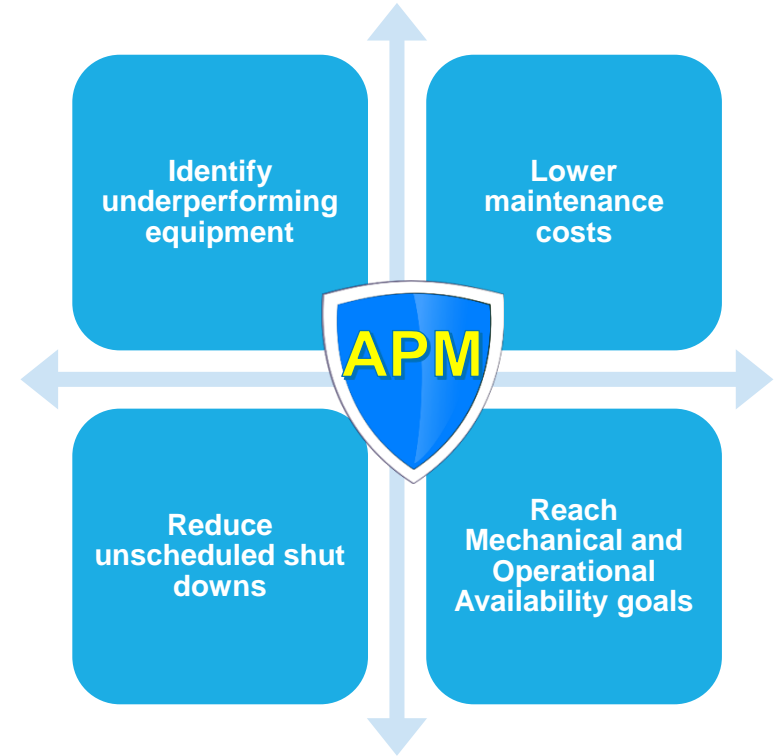




Proactive/Predictive Asset Analytics

Advanced Predictive Maintenance

- No Unplanned Downtime
- Advanced maintenance strategies leveraging PI AF, EF, and SAP PM Integration
- A new project plan was initiated in 2015 to establish a condition-based maintenance system via a new Integrator for Business Analytics applied to SAP PM



PI System & SAP PM Condition Based Maintenance (CBM)

PI System - SAP PM Integration

- SAP development
- PI System access via the Integrator for Business Analytics



Performance Based Maintenance Strategy

- PSA- Pressure swing absorbers
- Chillers
- Heat exchangers



Training & Support

- Knowledge sharing
- Exploration of further possibilities
- Continuous improvement

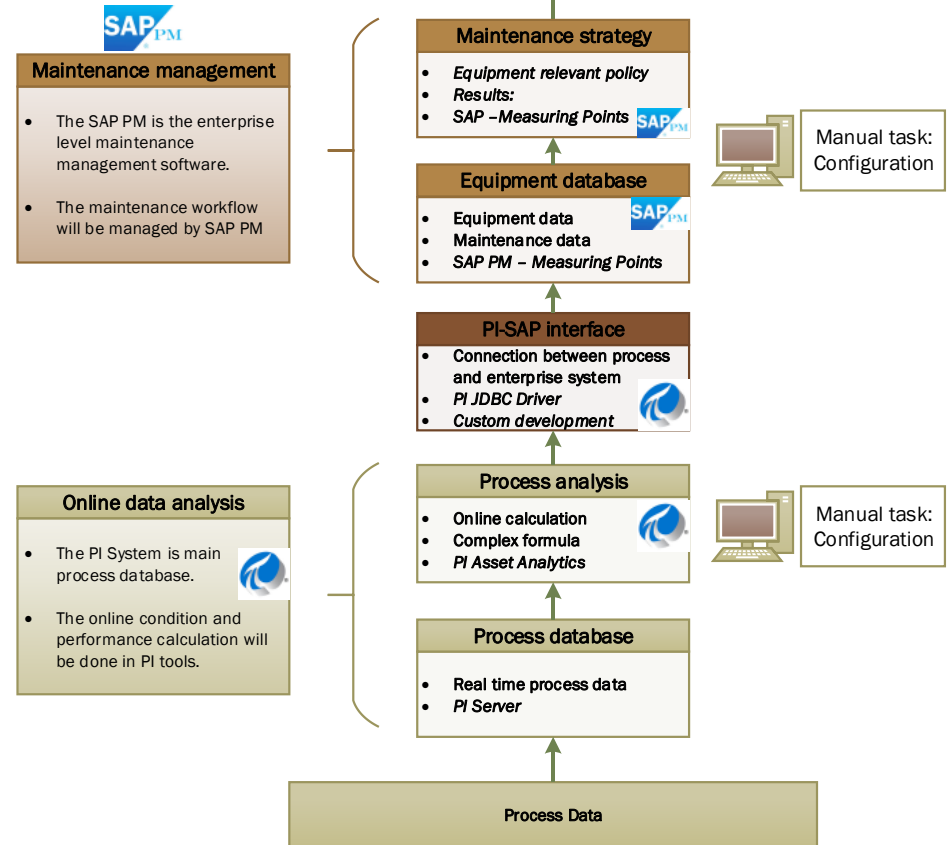


Performance or condition based strategy would fit more assets in the refinery than the time based. These calculations require process and operational data available in the PI Server.

The process data aggregation & **HealthScore calculation done in PI AF & sent to SAP PM.** Based on this result the Asset Maintenance Strategy is initiated by SAP PM.

Business Benefits

- Increased operational availability due to the installed strategies.
(PSA - Pressure Swing Absorbers in DHPP1-2, chillers in DSDW, heat exchangers where applicable)
- A working IT solution for further condition or performance based static asset policy implementation.
- KnowHow development about advanced condition based calculations.





Microsoft Azure Machine Learning

Goals and Scope

Proof of Concept project to utilize advanced analytics and machine learning in order to achieve similar results to APC in the following areas:

- Model for estimation and control of the sulfur content in diesel (Gas oil Hydrotreating unit)
- Optimization of the coke yields in the Delayed Coking Unit

Fully
managed

No software to install,
no hardware to manage,
and one portal to view
and update.

Integrated

Simple drag, drop and
connect interface for
Data Science. No need
for programming for
common tasks.

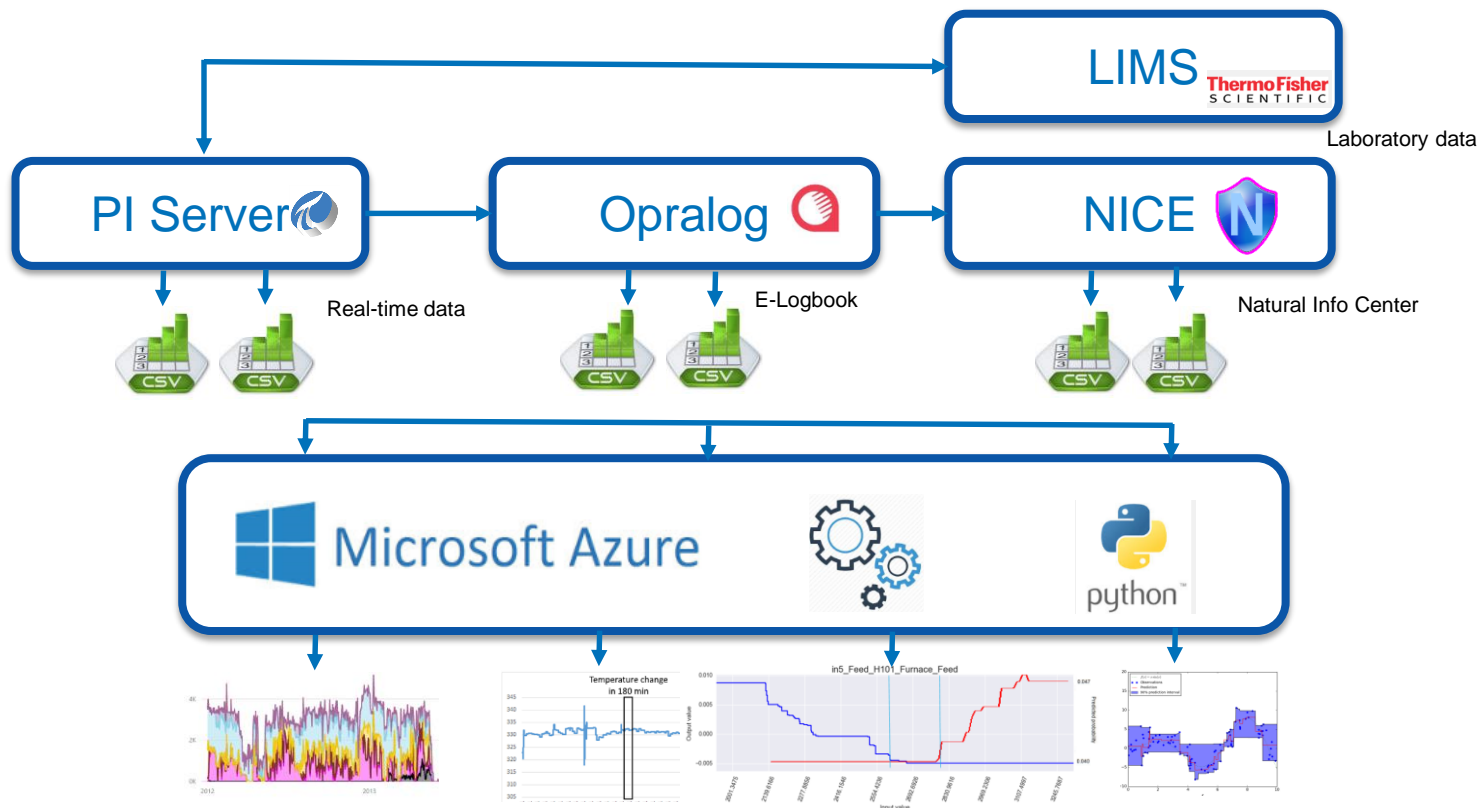
Best in Class
Algorithms

Built-in collection of best of
breed algorithms. Support for
R and Python for extensibility.

Deploy in
minutes

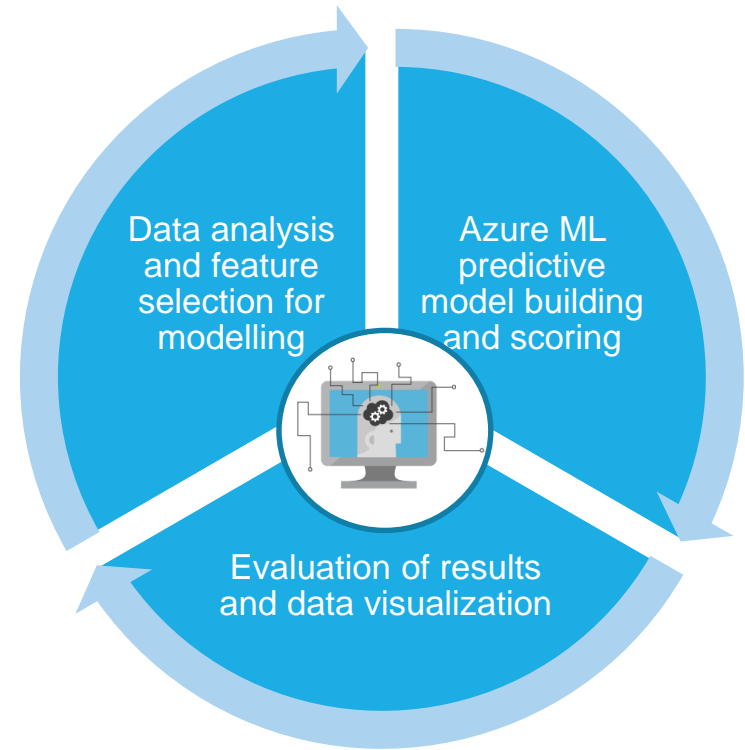
Operationalize models with a
single click. Monetize in
Machine Learning
Marketplace.


Machine Learning Architecture



Machine Learning- Use Cases

- Find the optimal mixture of different feeds into the Delayed Coking Unit (Residue Hydrocracking Unit & FCCU)
- Achieve minimal level of coke yield
- Gas Oil Hydrotreating Unit unit product sulfur content estimation based on available data
- Azure ML technology adaptation compare laboratory, online analyzer, APC and ML data

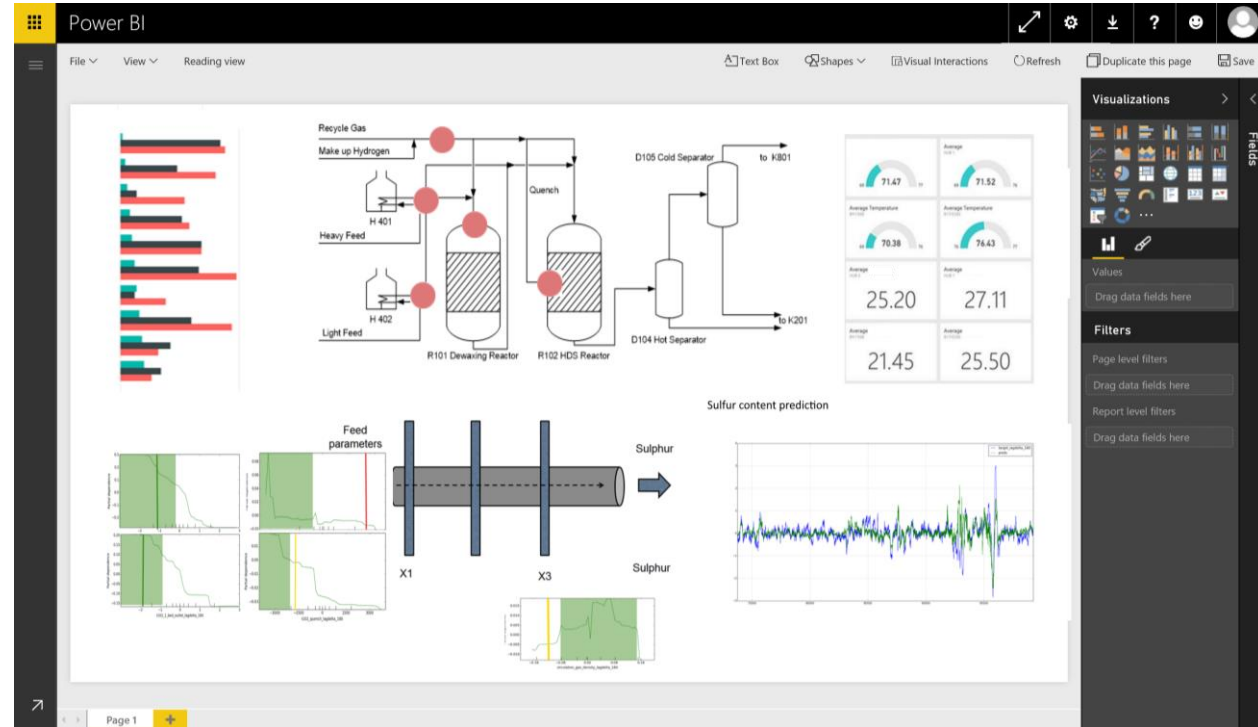




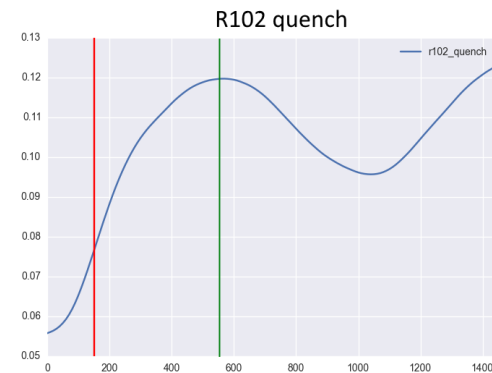
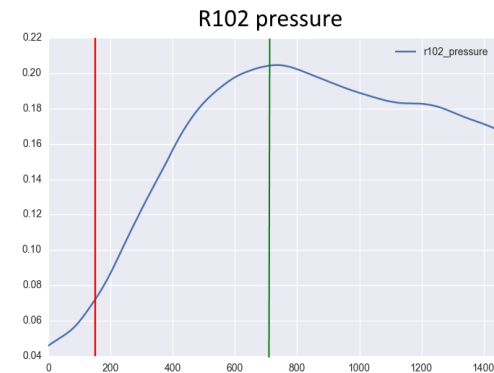
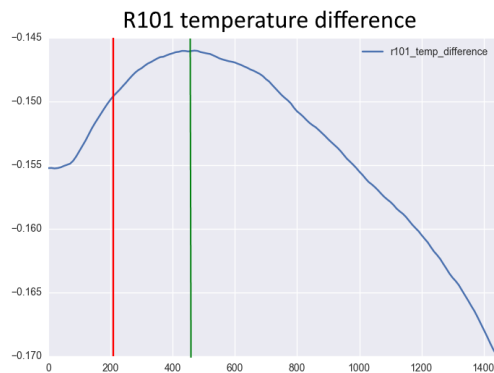
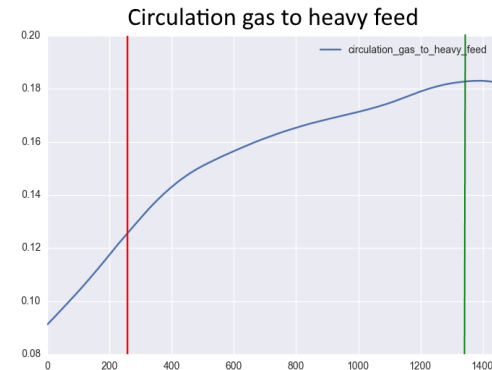
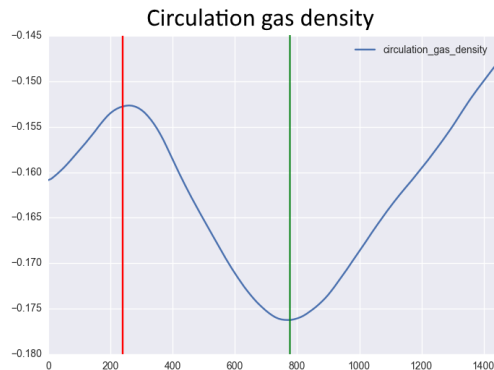
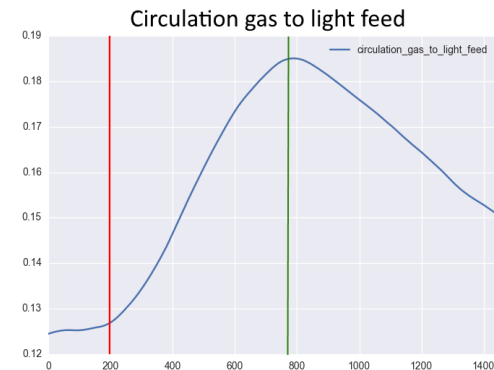
Predicting Diesel Sulfur as a in a Gas Oil Hydrotreating Unit

Project Scope – Rough Method to Optimize Sulfur in Diesel

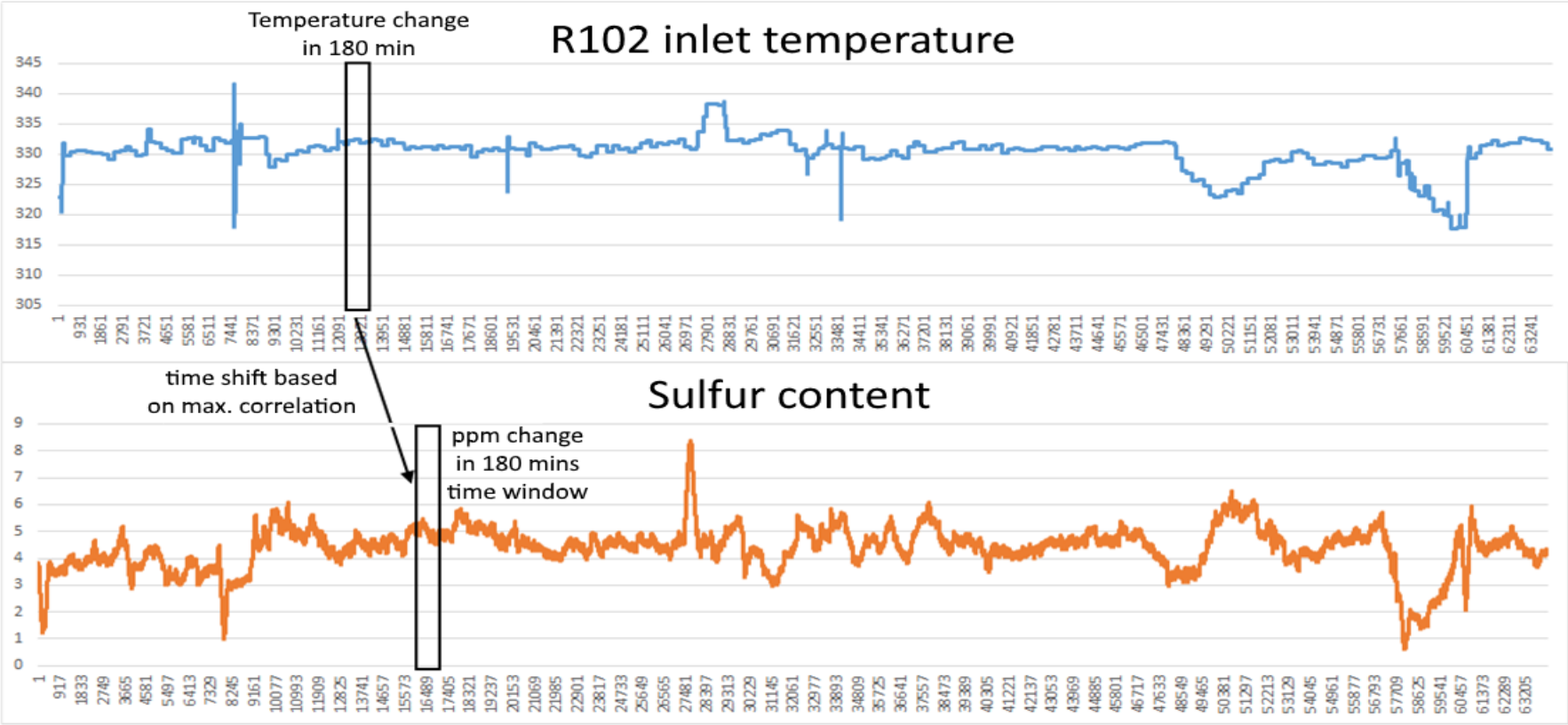
- I. Analysis of data to determine the control variables which has the highest effect on sulfur content in PI AF to Machine Learning and then back to PI AF
- II. Determine the desired intervals for each variable to control the output sulfur content
- III. Advanced model to (time window based) to predict the changes of the sulfur based on the changes in the control variables



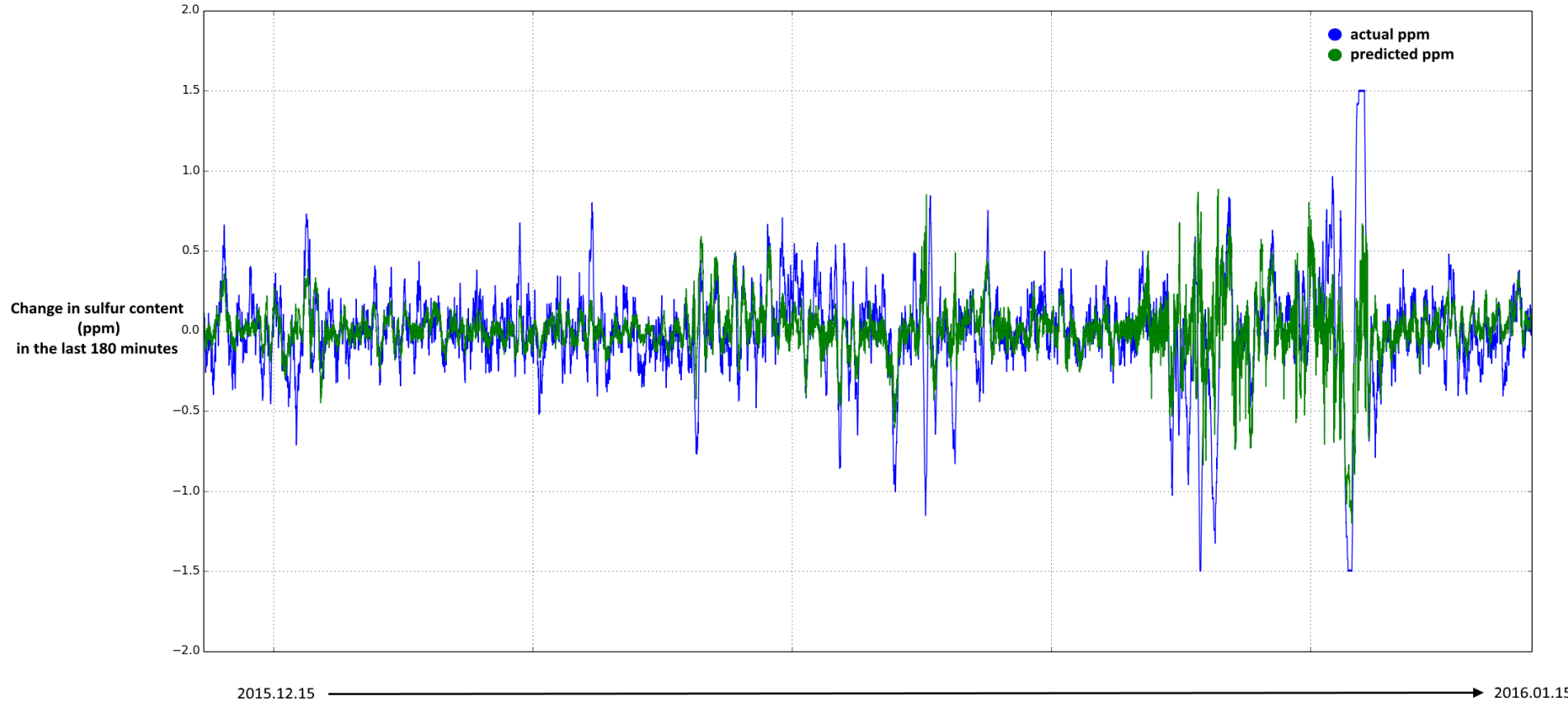
Time Shifted Correlation with Sulfur Content



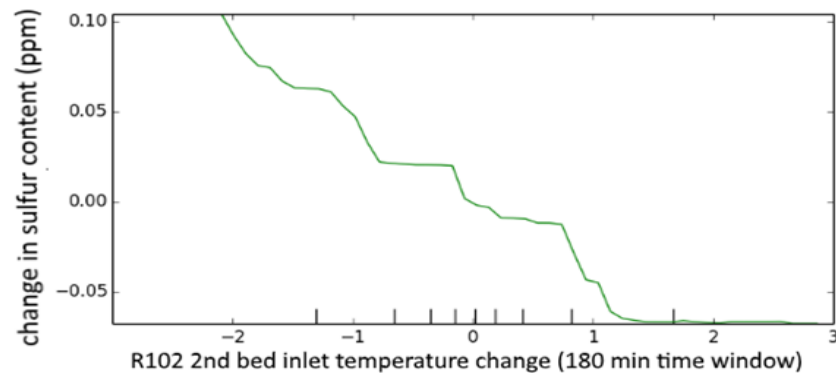
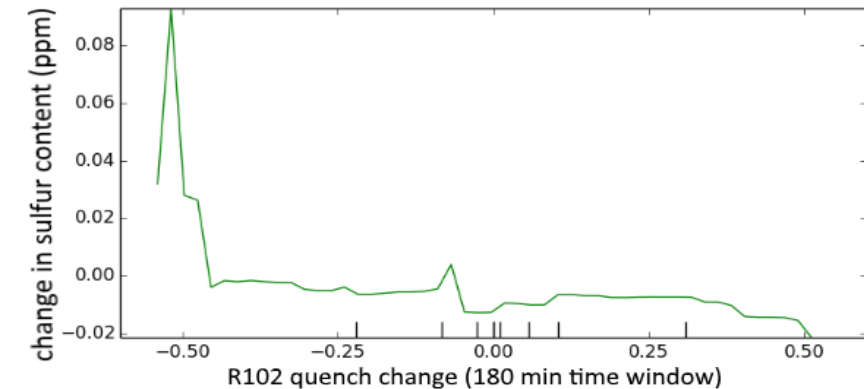
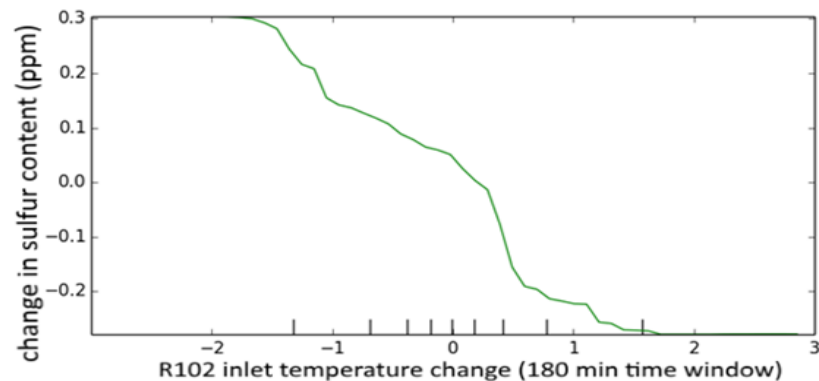
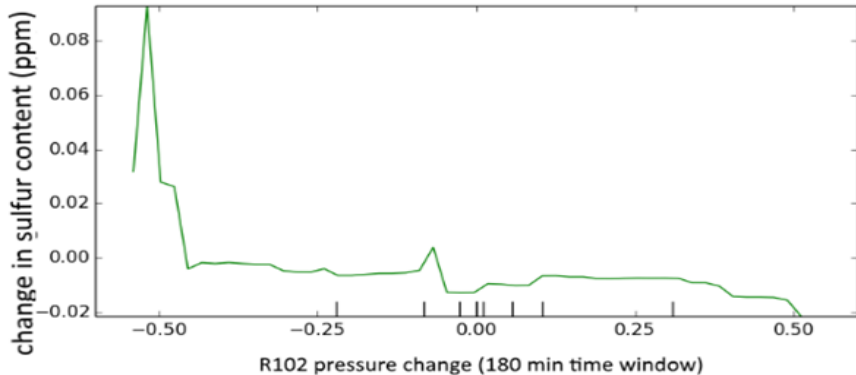
Predicting Output Based on Changes During a Time Period



Prediction and Control Model



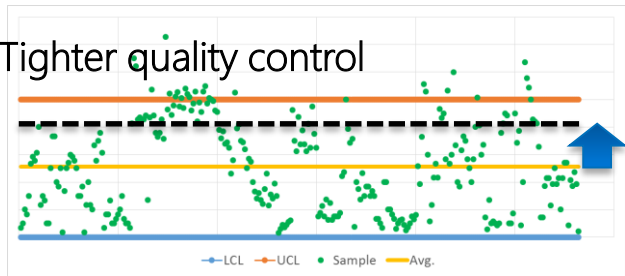
Influence of Sulfur Content



Benefits from Production Support

- Gas Oil Hydrotreater Unit #1 product sulfur content

- Tighter quality control



Quality giveaway decrease	3	ppm
Operation time (in feasible operation mode)	20	%
Fuel gas decrease	4495	GJ/Year
Benefits from decreased fuel gas consumption	47.000	\$/Year

- Effect of other empirical inferential

Maintenance of existing empirical inferential calculations
(DGHT3 Sulfur content, Cloud point, Pensky Martens FP, Gasoline EBP)

28.000 \$/Year

New empirical inferential calculations (E.g. NHT, Amine systems....)

70.000 \$/Year

- Sum of possible benefit ~ **\$140K/Year**



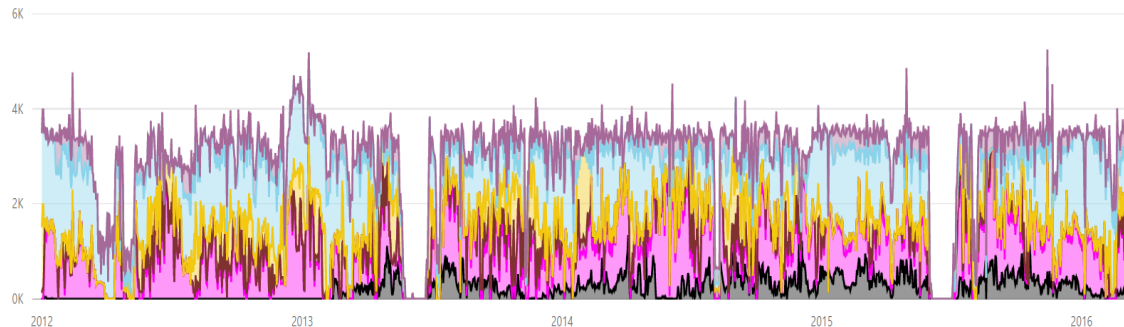
Predicting and Optimizing Coke Yield in a DCU

Azure Data Analysis Sample / Residue Hydrocracking Unit Flow to the Delayed Coking Unit

- We took coking cycles as the main aggregation unit (in PI AF)
- Aggregated the input hourly data to the cycles from **high fidelity operating data**
- Matched the daily coking cycle data in Machine Learning to coking cycles from PI Event Frames

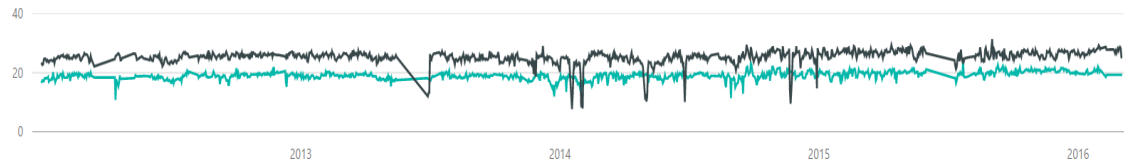
RHC_MASS, MIX_FROM_TANKS, DAV2Dark Distillate_MASS, DAV2GOUTRON_MASS, DAV3GOUTRON_MASS and propane deasphalting PEB_MASS by REFDATE

● RHC_MASS ● MIX_FROM_TANKS ● DAV2Dark Distillate_MASS ● DAV2GOUTRON_MASS ● DAV3GOUTRON_MASS ● propane deasphalting PEB_MASS



Average of AAG Conradson Carbon (m/m %) and Average of IND KOKSZ HOZAM SZAZALEK by REFDATE

● Average of AAG Conradson Carbon (m/m %) ● Average of IND KOKSZ HOZAM SZAZALEK



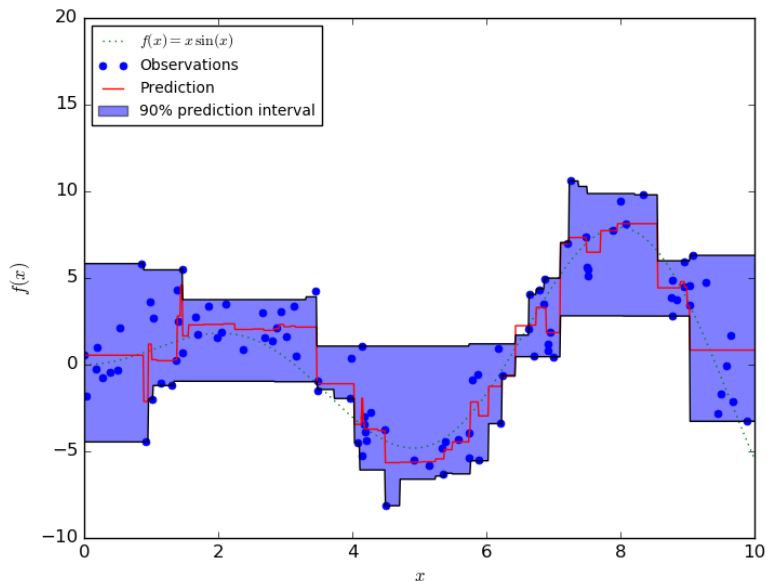
Average of Coke explosion value by REFDATE



Data from 2012-01-01 to 2016-03-01

Coke Yield Model

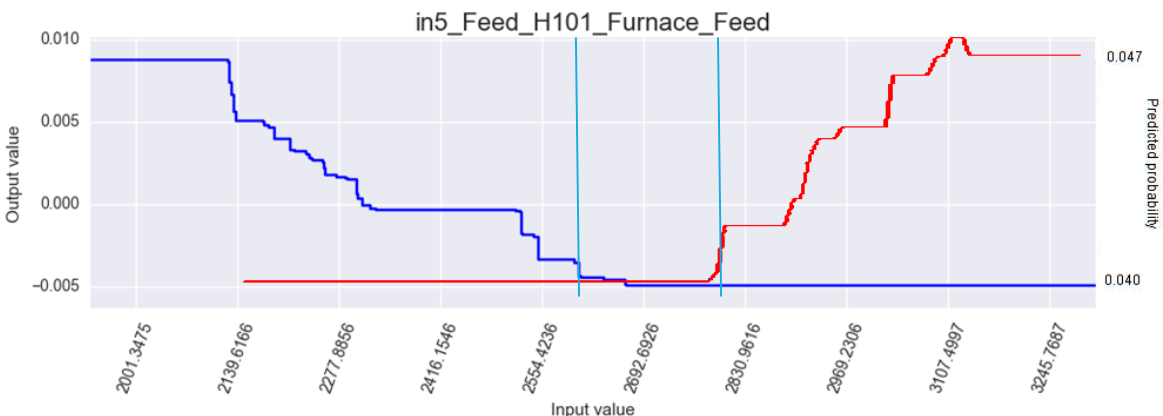
- Predicting the coke yield in percentage
- Discrete model, no time windows and time shifting
- GBR Algorithm (Gradient Boosting Regressor)



Timeline	Baseline (MSE)	Result (MSE)
2012-01-01 - 2014-12-31	0.000513	0.000363
2014-01-01 - 2016-03-03	0.000702	0.000366

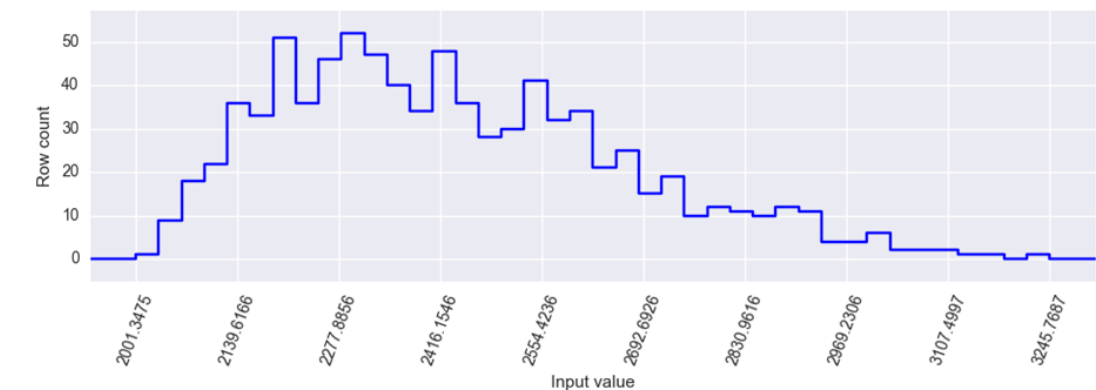
MSE: mean squared error

Coke Yield & Decoking Pressure Excursion Event



Blue: coke yield (output value)
Red: coke steam/hot pocket event likelihood

~In case of > 3100 t Furnace feed input the coke steam/hot pocket event likelihood is increasing

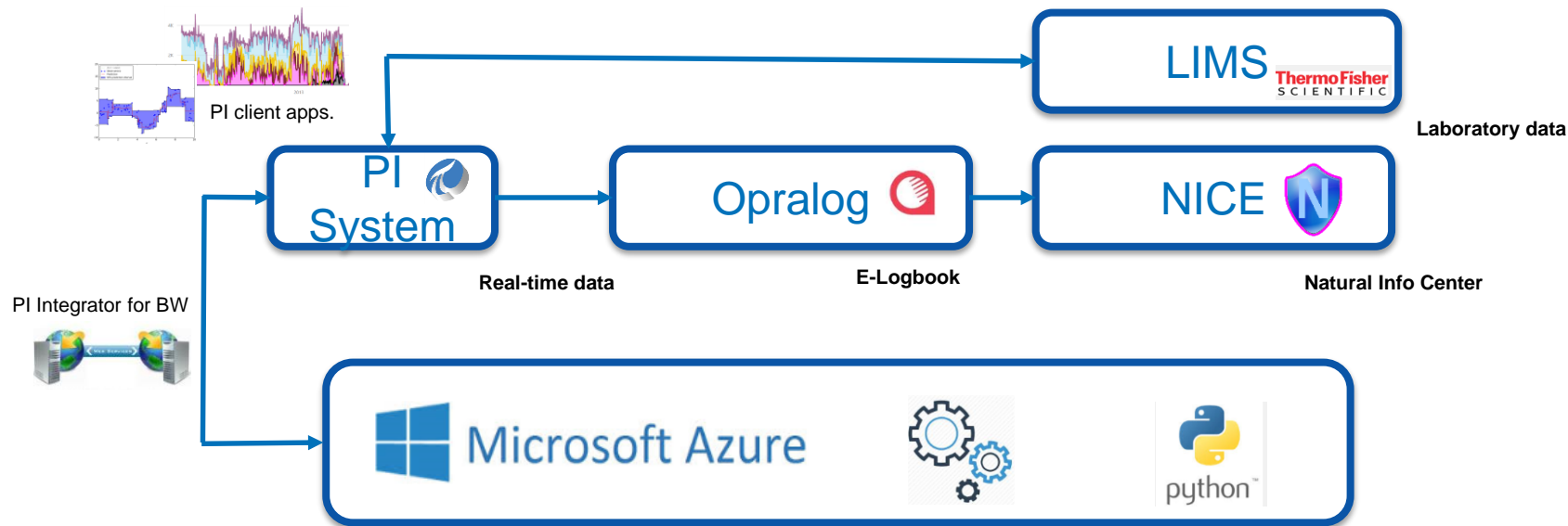


Blue histogram:
Row count coke cycle

Between the 2550 - 2800 t intervallic the coke yield could be decreased without coke steam/hot pocket event

Event Frames is the monitoring tool

Machine Learning architecture – Future Plans



MOL IT Security approval required

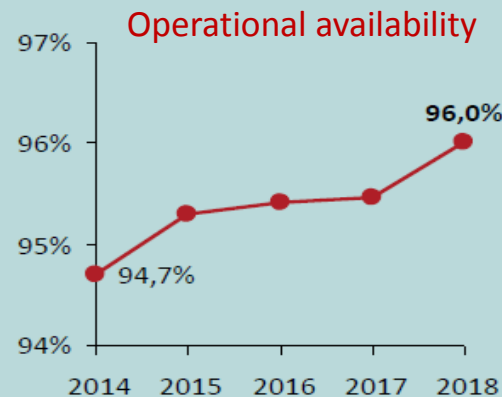


Conclusion

PI AF is a Powerful Tool to Deliver Business Value from Advanced Analytics

COMPANY and GOAL

Deliver an Incremental \$500M in EBITDA from the PI System & PI AF Enabled Improvement Programs Leveraging Advanced Analytics & Machine Learning to Support IOW, Forecasting, and Proactive/Preventative Maintenance



CHALLENGE

Aggressive New Downstream and NeXt Downstream Program 2018 Objectives; Limited CapX and OpX:

- Asset Reliability and Integrity
- ISO50001 Energy Forecasting
- Process Alternative crude processing and optimize results in different yield structure

SOLUTION

Continue to Leverage PI AF to perform Operating Windows, Forecasting, and Machine Learning with MS Azure:

- Standard & Critical Integrity Operating Windows
- Corrosion Advanced analytics
- Hourly NG forecasting
- DGHT3 Diesel Sulfur Prediction
- DCU Coke Make Optimization

RESULTS

Supporting \$500M EBITDA New Downstream Program 2018 Goals - Examples:

- 1-2.0% reduction in PSA Pressure Swing Absorbers breakdown in Hydrogen production plant: \$200K/yr
- Increased Mechanical and Operational availability to 96.0 worth >\$100M/yr

Contact Information

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Process information and
Automation leader

MOL Plc.



Questions

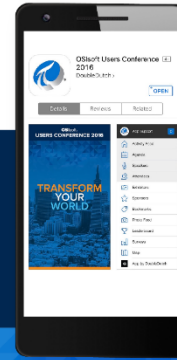
Please wait for the **microphone** before asking your questions



State your **name & company**

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谢谢

Danke

Merci

Gracias

Thank You

ありがとう

Köszönöm

Спасибо

Obrigado

The background of the slide is a dark blue gradient with a faint, stylized image of the San Francisco skyline, including the Golden Gate Bridge and the Transamerica Pyramid. The OSIsoft logo is positioned at the top center.

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