



Real-time Estimations and Online Learning for Industrial Assets at Total

Presented by **SPINNATO Juliette**



AGENDA

- ▶ TOTAL: our Company and our data-driven philosophy
- ▶ From a Business use case to a data science “Proof Of Concept”
- ▶ Online learning for our industrial assets
- ▶ Conclusion and Next steps



TOTAL: our Company and our data-driven philosophy

TOTAL: Committed to Better Energy

Total is the world's **4th**-ranked oil
and gas company¹
and a global **leader** in solar energy
with SunPower.

WITH OPERATIONS IN MORE THAN

130 COUNTRIES,

we have over **96,000 employees**
who are fully committed to better energy.

¹ Based on market capitalization in U.S. dollars at December 31, 2015

TOTAL: Committed to Better Energy

TOTAL'S ACTIVITIES

EXPLORE AND PRODUCE

- ① OIL AND GAS
- ② SOLAR
- ③ BIOMASS

TRANSFORM AND DEVELOP

- ④ SPECIALTY CHEMICALS
- ⑤ POLYMERS
- ⑥ REFINING - PETROCHEMICALS

SHIP AND MARKET

- ⑦ TRADING - SHIPPING
- ⑧ MARKETING AND SERVICES



TOTAL: Our Data-Driven philosophy

A DATA-DRIVEN ACTIVITY WITHIN TOTAL

- ▶ Since 2015, Total promotes internally **a strong data culture**
- ▶ To move **towards a data-driven and digitalized industry**
- ▶ Creation of **data analytics centers** within our company
 - Realization of several Proof of Concepts (PoC) to demonstrate **data science added-value for our Businesses**
 - Industrialization of the best PoCs into long-term solutions



From a Business use case to a data science POC

USE CASE



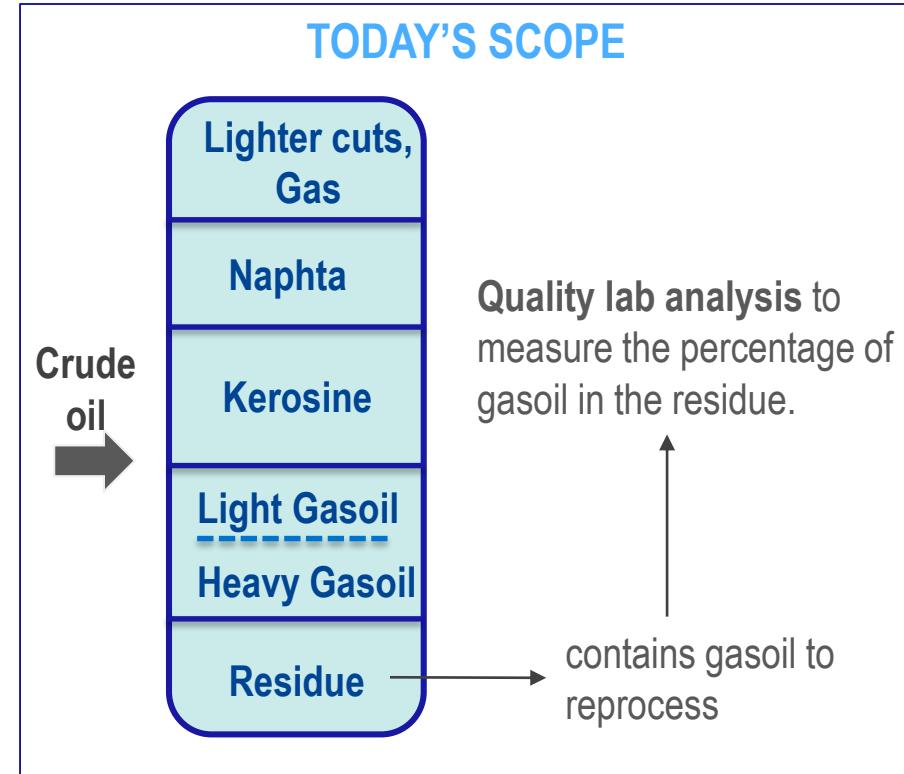
BUSINESS OBJECTIVE

Increase lab & units efficiency using advanced analytics

- ▶ To reduce the amount of samples processed that bring little or no added value
- ▶ To help the process experts to improve units outputs



A proof of concept: estimating crude distillation quality for Antwerp refinery



A DATA SCIENCE PROOF OF CONCEPT

SUCCESSFUL METHODOLOGY

- ▶ A 5 months project following an AGILE method
- ▶ A multi-disciplinary team including Business experts, data scientist, and data owner

DATA SOURCE

80 relevant PI Tags™ identified and extracted from the PI System®

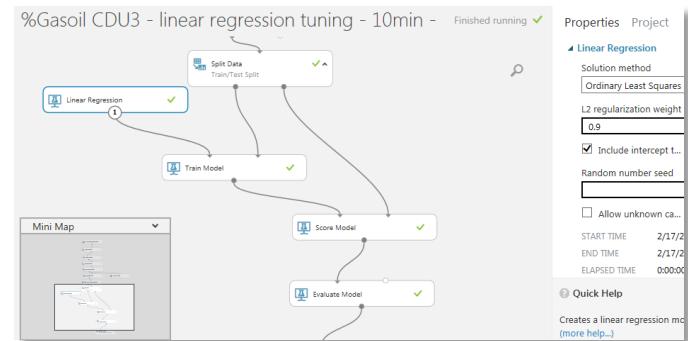
A one shot study: **data extraction using PI DataLink®**

MACHINE LEARNING MODELING

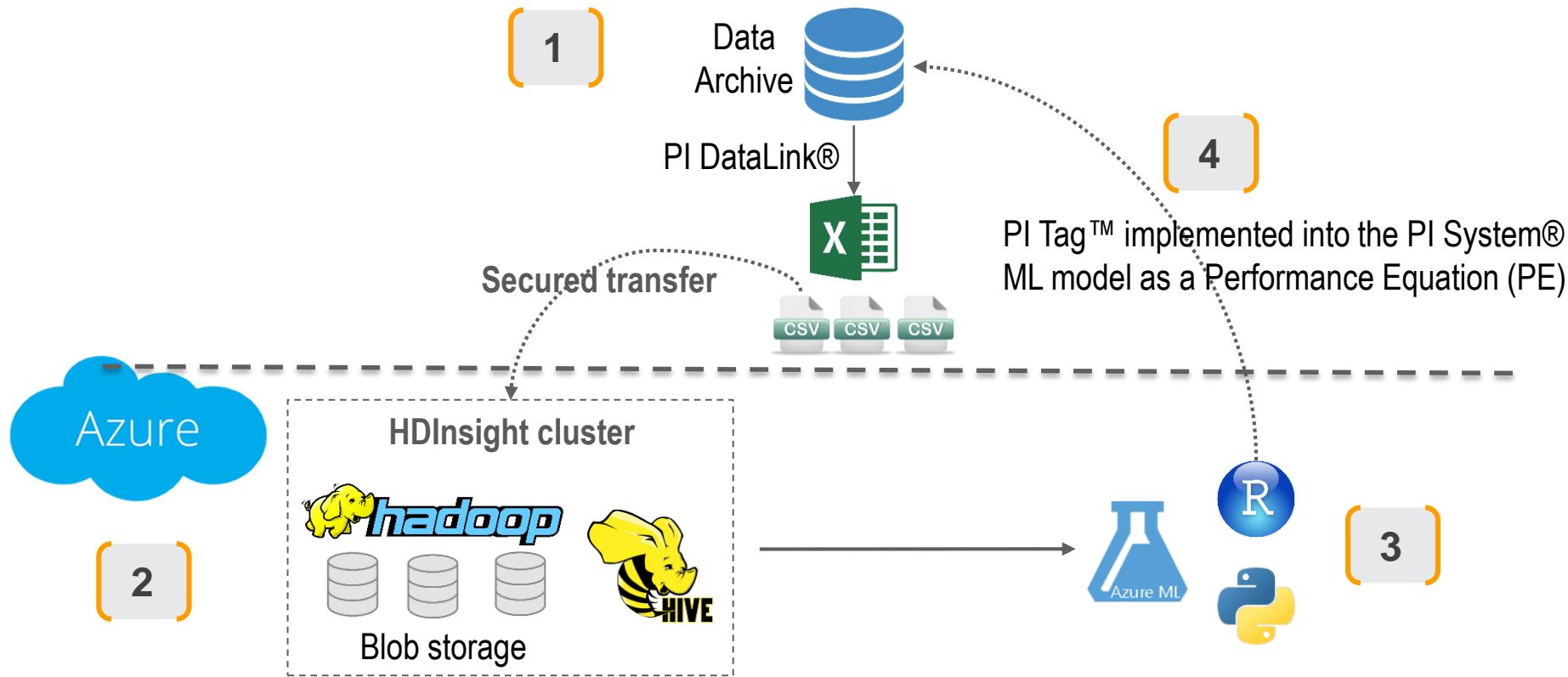
- ▶ White box machine learning algorithm
- ▶ Highly correlated variables (flow, pressures, temperatures, ...)

Linear Ridge regression

Azure ML Studio:



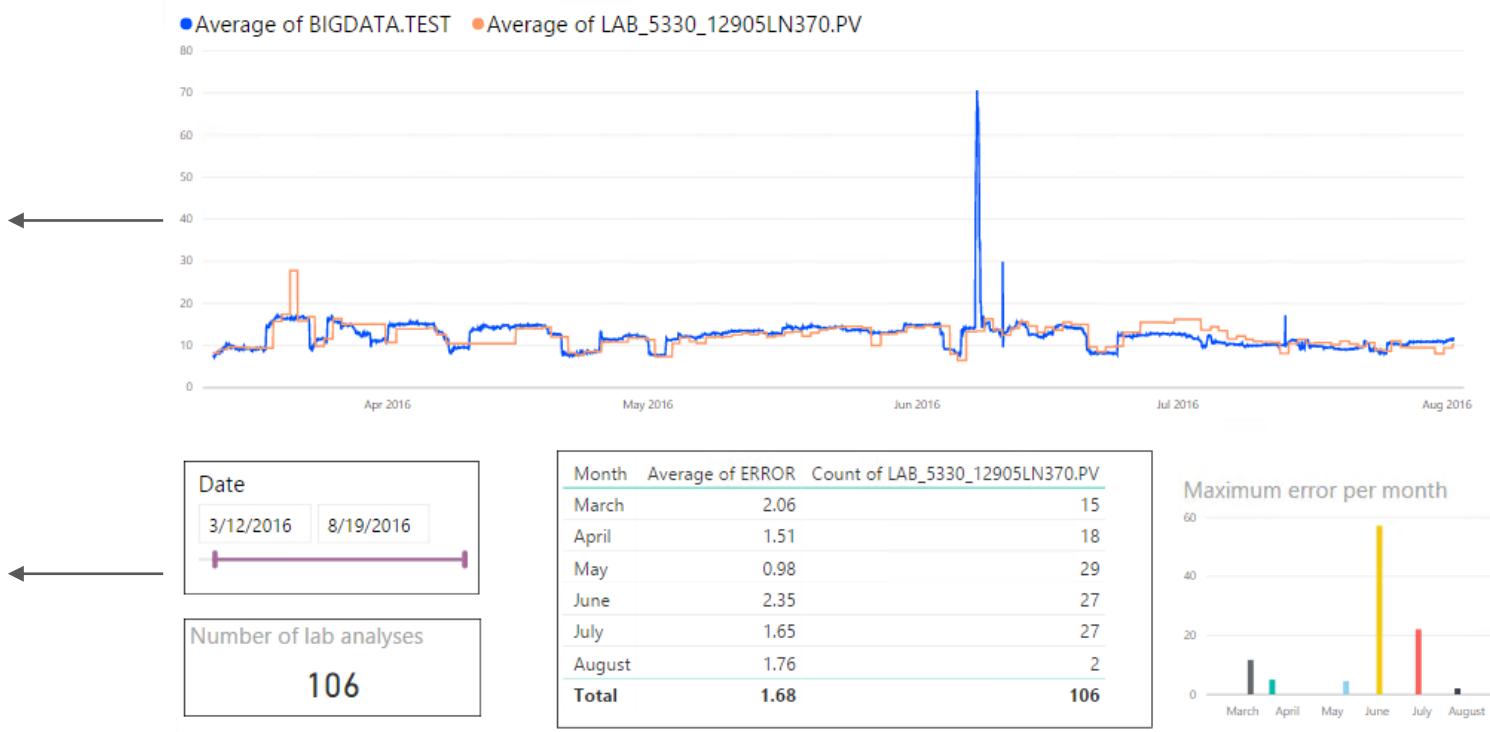
ARCHITECTURE & WORKFLOW



RESULTS

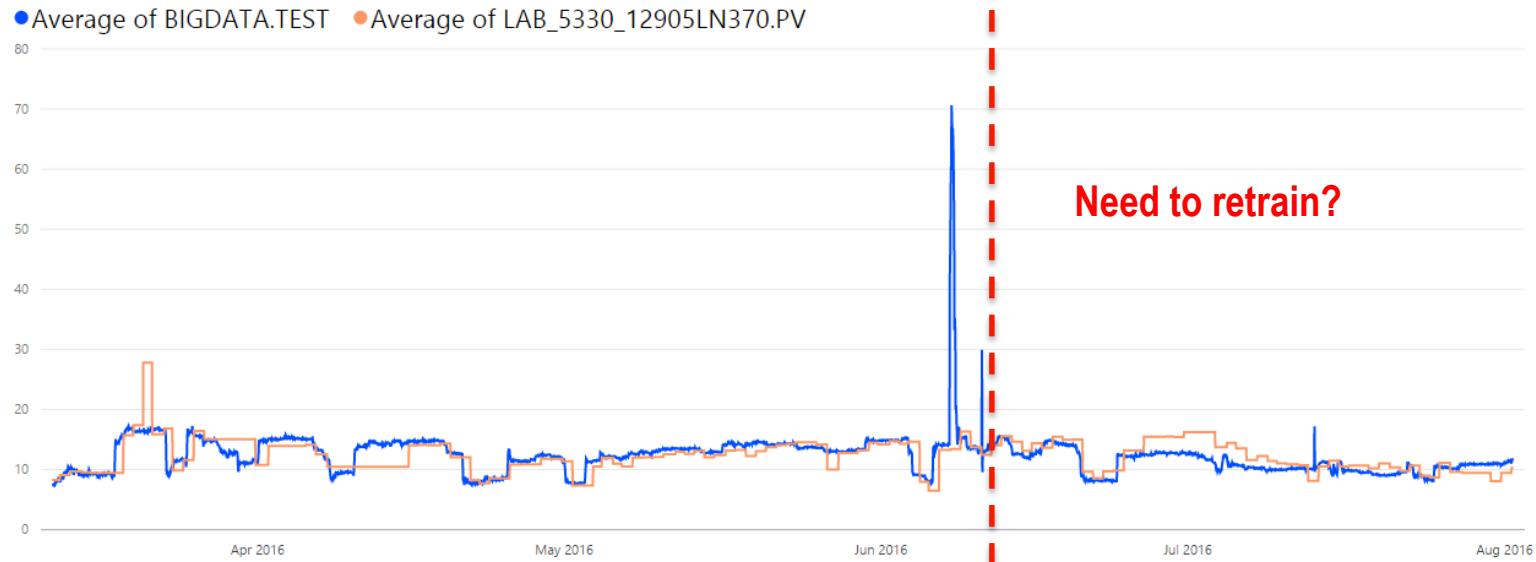
REAL-TIME ESTIMATION AND MODEL FOLLOW-UP

Model estimation
versus
lab analysis



RESULTS

MODEL SUSTAINABILITY



- ▶ When do we decide that the model is no longer up-to-date?
- ▶ When do we decide to retrain the model?

CHALLENGES

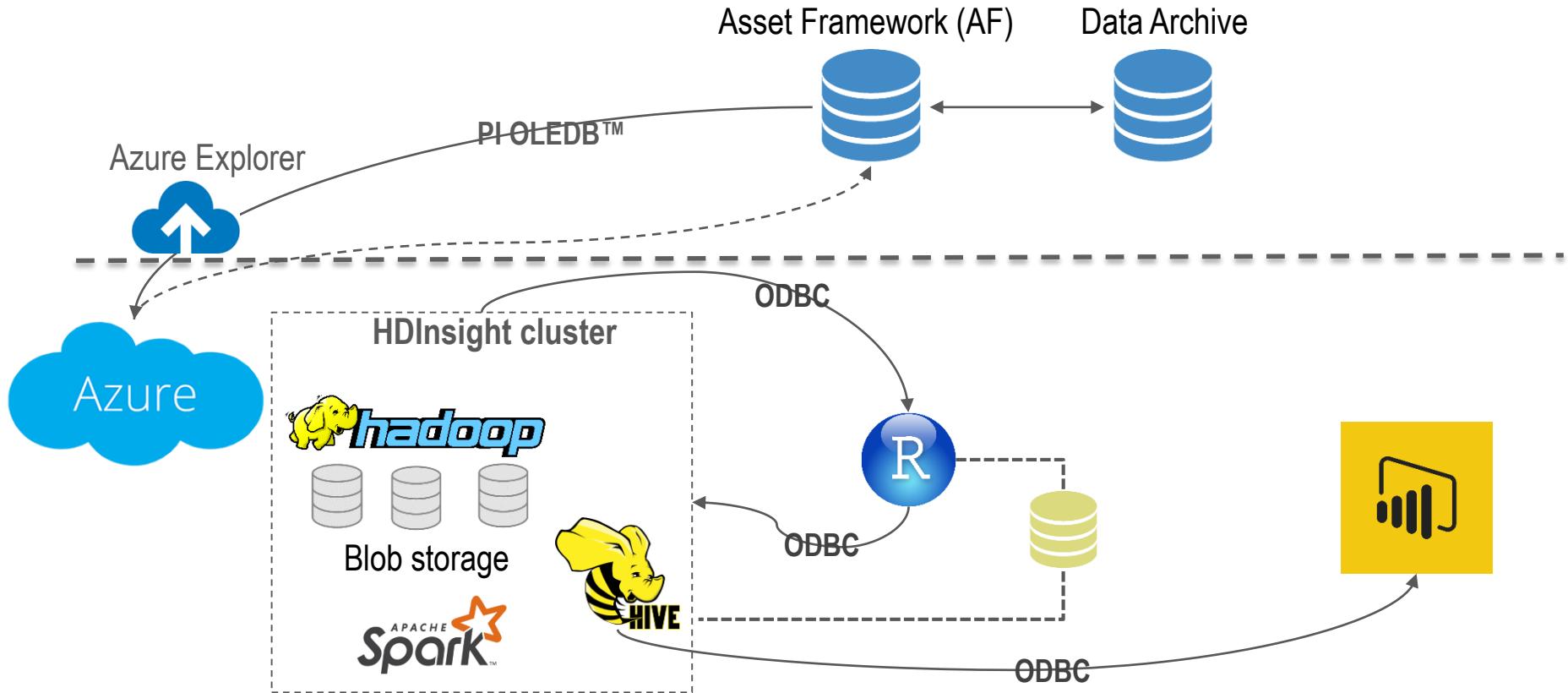
Maintaining a ML model implemented in a PI System® : KEY QUESTIONS

- ▶ How do we **decide to retrain** the model: detect changes made on the industrial asset, launch retraining based on a statistical and/or business criteria
- ▶ What is **the best dataset** to take into account?
- ▶ How **to keep the historic** of our online learning?

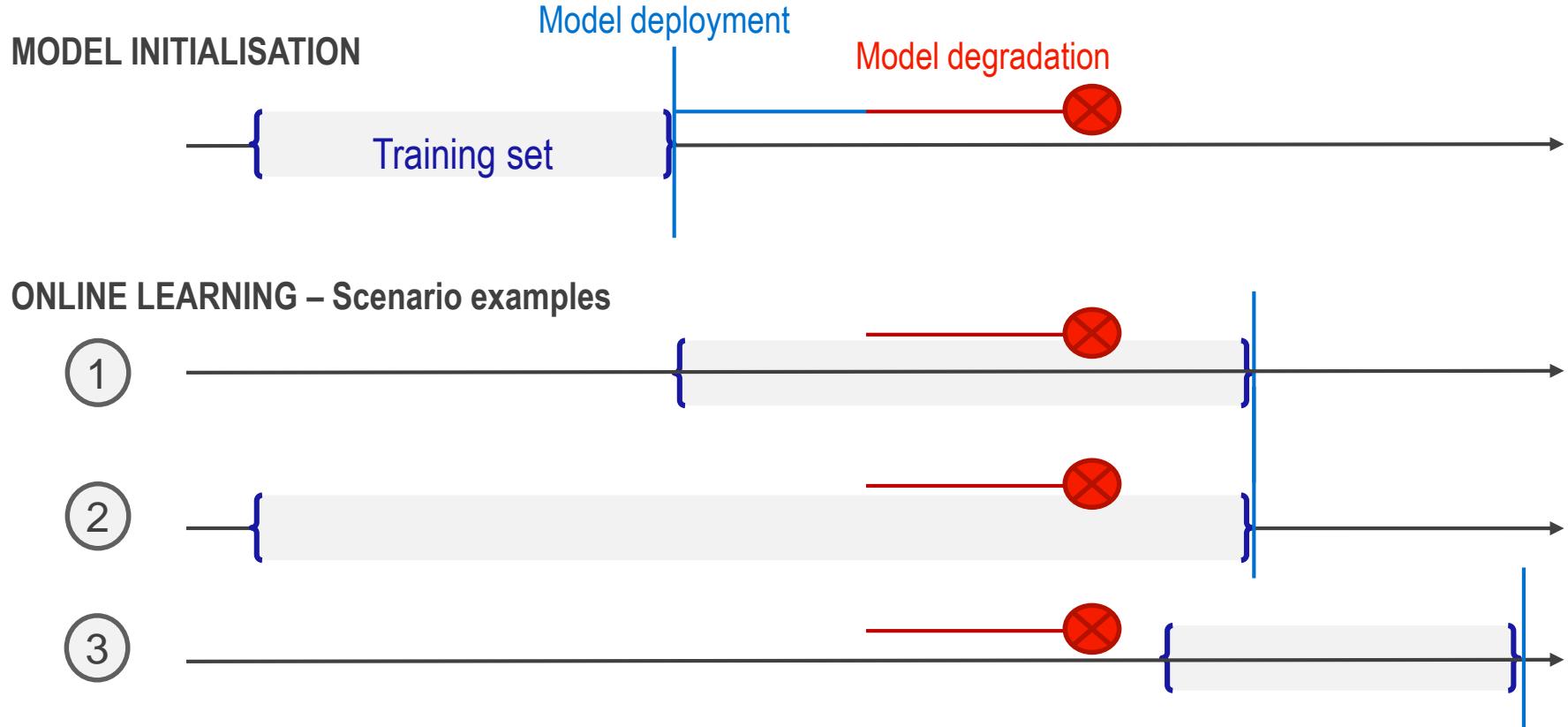


Online learning for our industrial assets

TARGET ARCHITECTURE



ONLINE LEARNING: Dataset(s) & model(s)



ONLINE LEARNING: Expert System

How to...

- ▶ ...automatically define the different datasets and
- ▶ ...evaluate models



Start_Control_Algorithm ()

Inter_V: The time interval controls tags = 30 minutes

Inter_M: Variable check interval = 10 minutes

Perf_model = F (Mean error, accuracy, actual deviation)

Perf_threshold : Performance threshold set by Business

For each Inter_M

If (Perf_model < Seuil_Perf) *then*

Create N datasets

Dataset = {old dataset, Delta-2, Delta-1, Delta 1, Delta2 }

Modeles_test <- Create 2^N prediction models

Model <- Max (Modeles_test) # the best model of Modeles_test

End if

End for

End_Control_Algorithm

ONLINE LEARNING: First results

MODELING CHOICES

- ▶ Elastic-net algorithm: Variables selection and optimized linear regression



R library : “***glmnet: Lasso and Elastic-Net Regularized Generalized Linear Models***”

Ref: [T. Hastie et al. , “*The Elements of Statistical Learning*”, Springer Series in Statistics, 2nd Edition, Springer, 2009]

MODELS EVALUATION

- ▶ **Performance evaluation:** Mean-absolute error (MAE)
- ▶ **Business evaluation:**

Model equation highlights the main contributions to estimate the target

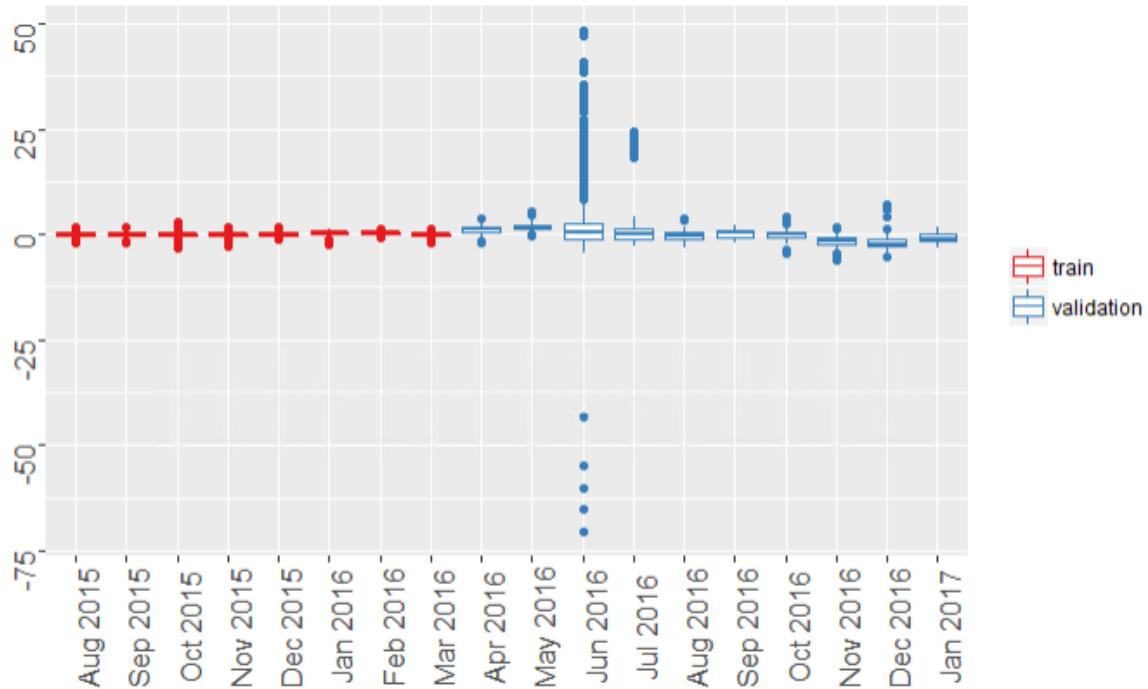
Descriptive statistics



ONLINE LEARNING: First results

R RESULTS - Example

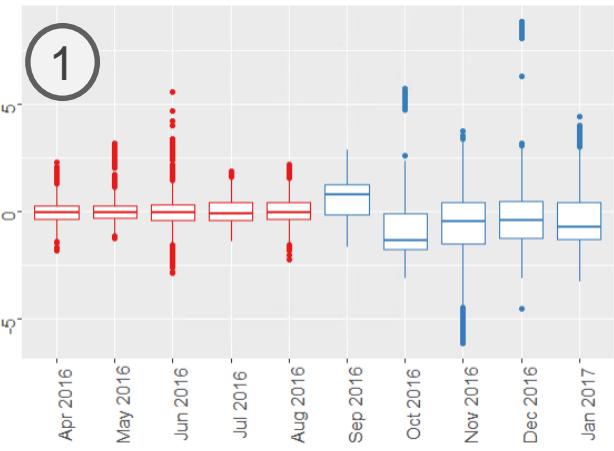
MODEL INITIALISATION – Estimation Error



	Training	Validation
MAE	0.36	2.08

ONLINE LEARNING: First results

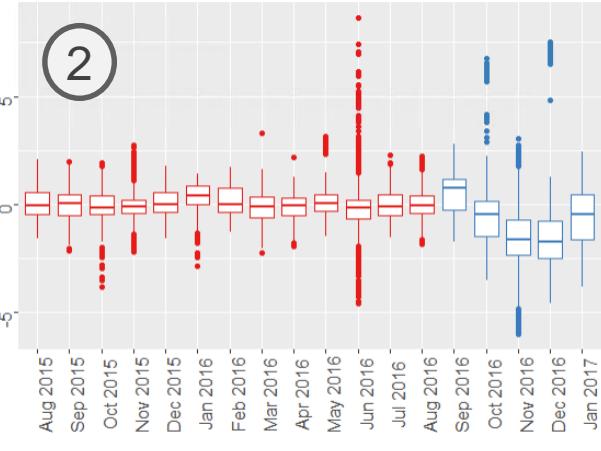
R RESULTS - Example



Training Validation

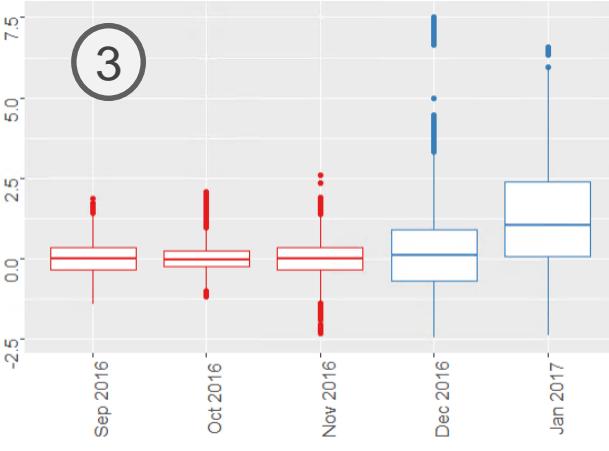
MAE

1.32 0.93



Training Validation

0.53 1.38



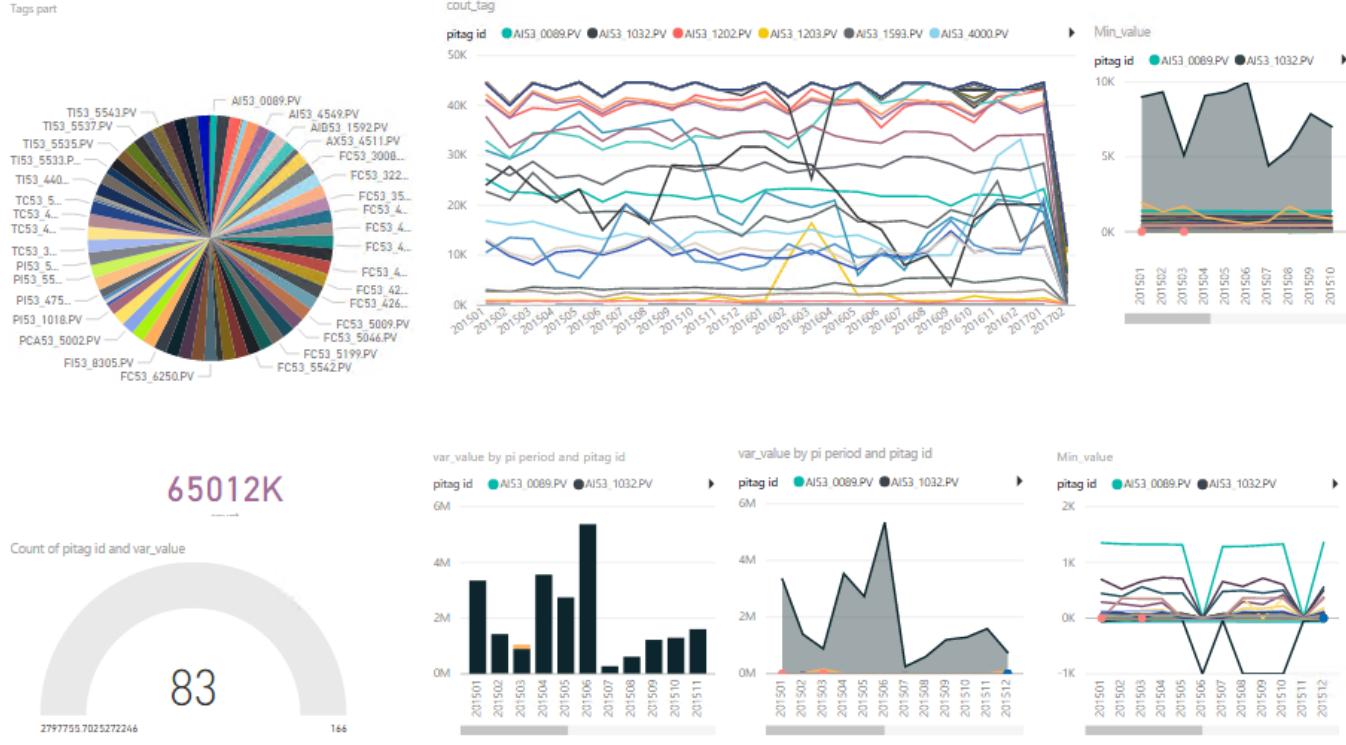
Training Validation

2.52 1.52

ONLINE LEARNING: Models follow-up

Power BI Dashboard

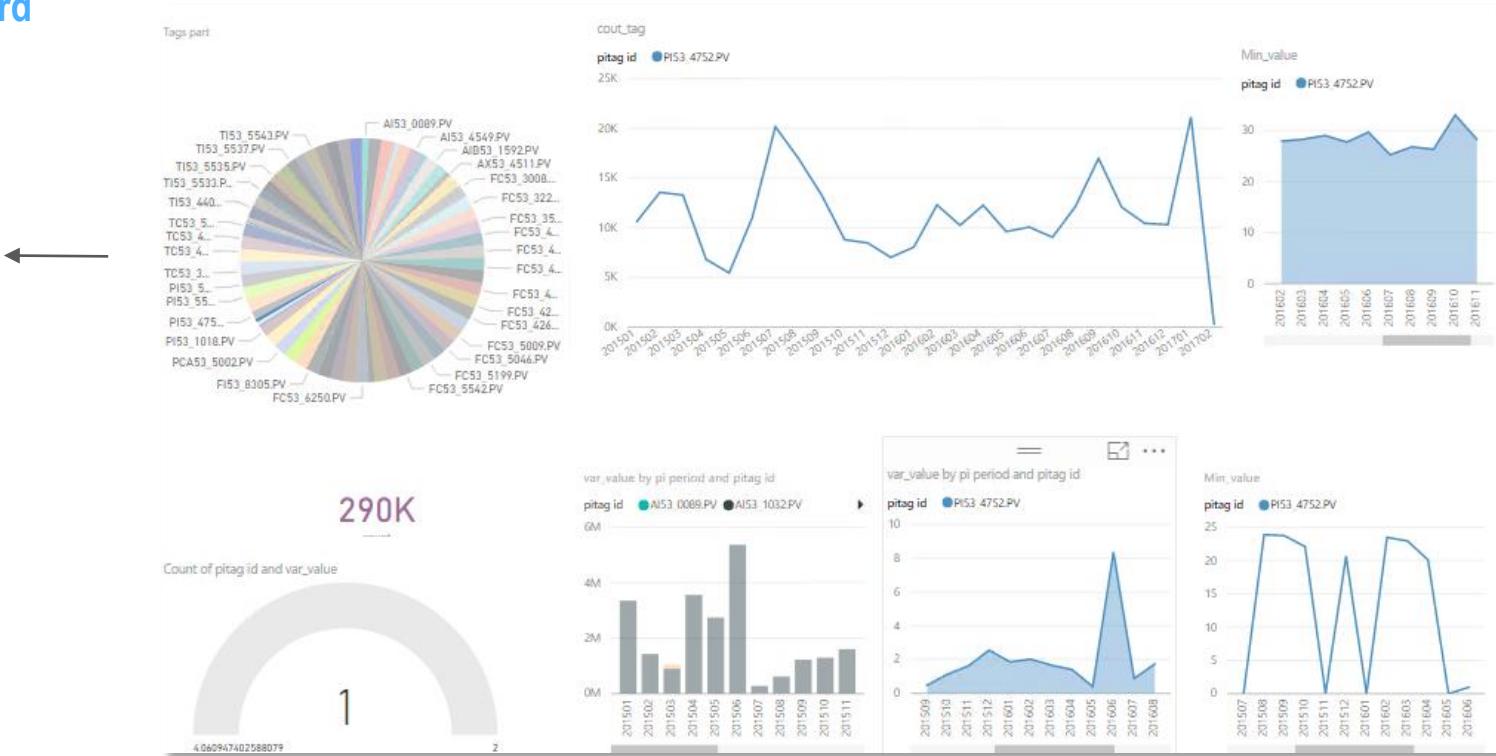
PI Tags™
descriptive statistics
to detect important
changes among
input variables



ONLINE LEARNING: Models follow-up

Power BI Dashboard

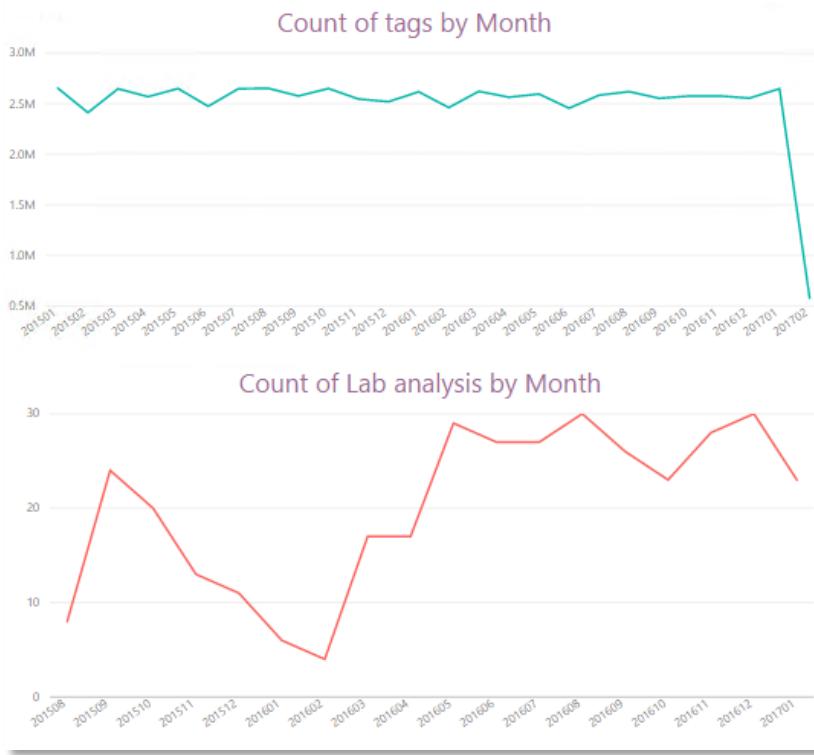
Dynamic dashboard



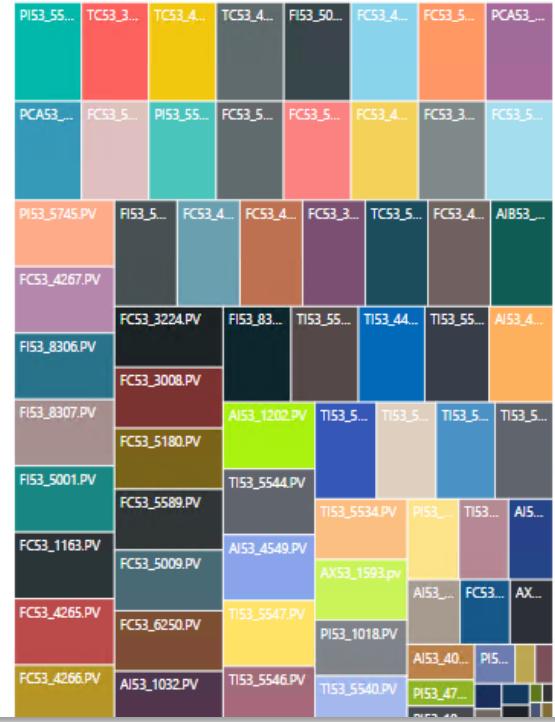
ONLINE LEARNING: Models follow-up

Power BI Dashboard

Data quantity to monitor models retraining

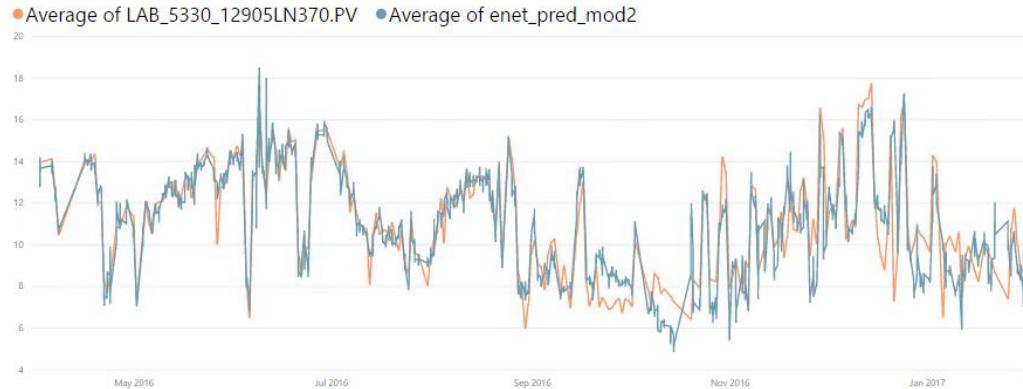


Select a tag



ONLINE LEARNING: Models follow-up

Power BI Dashboard



Model performances in time

Variables influence

PI53_5532.PV
PCAS3_5002.PV
AI53_4546.PV
PI53_1016.PV
AI53_4549.PV
AI53_1203.PV
AX53_4511.PV
TI53_5535.PV
TI53_4409.PV
PI53_4752.PV
TI53_5547.PV
TI53_5536.PV
TI53_0070.PV
AI53_5006.PV
TI53_5537.PV
FC53_4265.PV
FC53_5200.PV
TI53_5249.PV
FC53_4264.PV
FC53_3008.PV
TI53_4504.PV

Variables influence for each model



Conclusion and Next steps

CONCLUSIONS

BUSINESS OBJECTIVE

Increase lab & units efficiency using advanced analytics

DATA SCIENCE SOLUTION

Online learning to estimate the % of gasoil remaining in the residue of a distillation unit

- ▶ An automated workflow to extract data, compute several models and provide data scientists and Business experts with an updated dashboard

ADDED VALUE

Short term impact

- ▶ The proposed workflow is easy to both install and generalize many ML models for quality estimation
- ▶ Provide a clear and updated view of unit's quality to the Business at any time

Long term impact

- ▶ Important reduction of processed samples that bring little or no added value

NEXT STEPS

- ▶ Create more rules to launch **a model retraining automatically**
- ▶ Give Business experts **the possibility to manually launch a model retraining**
- ▶ Send online results back to the refinery's Asset Framework (AF)

Real-time Estimations and Online Learning for Industrial Assets at Total

COMPANY and GOAL

TOTAL: committed to better energy

The Business objective is to increase lab & units efficiency using data science methodology and tools



CHALLENGE

To build a data-driven model based on PI Tags™

In order to anticipate possible model degradation over time, **an online retraining** has also been developed.

- A project for Antwerp Refinery (Belgium) to estimate the % of gasoil in the residue of a Crude Distillation Unit.

SOLUTION

Definition of an online machine learning workflow using data science and PI systems tools

- Asset Framework (AF) and PI OLEDB™ were used to efficiently extract data
- Data science tools



RESULTS

An automated workflow scalable, easy to use and that provides good quality estimation

- Model estimation providing an error of less than 1.5 point
- A clear and updated view of unit's quality to the Business at any time
- Lab samples reduction

Contact Information

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Total



Questions

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



State your **name & company**

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감사합니다

Merci

Danke

谢谢

Gracias

ありがとうございます

Спасибо

Obrigado

Thank You