



Machine learning soft-sensors trained with Digital Twin for improving product quality & reducing energy

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AGENDA

- Intro to CEPSA
- The plant revamp and the OTS
- Intro to Inprocess
- Generating datasets from HYSYS
- Creating the soft sensors
- Results and conclusions



About CEPSA



THE CARLYLE GROUP



Cepsa is a **global**, **integrated** company operating across **the entire oil and gas value chain** and with over 90 years of experience.

Chemicals is one of the engines that is driving our internationalization and one of the areas where we are growing the most.

We are the world leading producer in Alkylbenzene (600 kMT/y total capacity)

Three LAB* Production sites located in San Roque (Spain), Becancour (Canada) and Camaçari (Brazil)



10,000 PROFESSIONALS

Who work with technical excellence, an ability to adapt, and an innovative spirit. They help us to stay competitive and are the key to face and overcome future challenges.



EXPERIENCE

We are supported by 90 years' experience in the oil world, making us a leading and robust company.



CONTINENTS

We operate across the entire oil value chain in five continents through our businesses in Exploration and Production, Chemicals, Refining, Distribution and Marketing, Gas & Power, and Trading.



INTERNATIONAL OUTLOOK

Our leadership in chemicals, combined with our broad experience in project execution across the world, helps us to grow internationally.



INTEGRATED BUSINESS

The physical integration of our production plants strengthens our model, reduces logistical costs and increases synergies.



SUSTAINABLE ENERGY

Our main priorities are to provide society with a safe, reliable and sustainable energy and contribute to the economic and social development of the communities where we work.



CUSTOMER SATISFACTION

Our commitment to the quality of our products and services and to the satisfaction of consumers forms the base of our customer relationship.



CHEMICALS LEADERSHIP

We are world leaders in LAB production (the raw material used to make biodegradable detergents), and in cumene. We are also the second world producer of phenol, used to make high performance materials, and acetone.



Innovation underpins everything we do. We have a Research Center that helps us to create value, optimize processes and improve the quality of operations and products.

CEPSA is an integrated oil & gas Company and the largest Linear Alkylbenzene producer
in the World (600 kMT/y)

Puente Mayorga Petrochemical Plant produces LAB, raw material for biodegradable detergents. It supplies around 50% of the African and 25% of Western Europe markets



250 kt/y LAB

80 kt/y LABSA

400 kt/y Paraffins

100 kt/y Solvents

- Began operations in 1969 to produce LAB, raw material for Biodegradable Detergents
- Located at Algeciras Bay
- First plant ever to retrofit to Detal Technology (2021)
- Betting on quality: First company in Spain to achieve ISO 9001 certification (year 1992)



The capacity expansion and technology revamp

• Cepsa implemented a new upgrade project to expand production at its Puente Mayorga Chemical Plant in San Roque (Cadiz). The revamping process covers the installation of the new Detal technology, co-licensed by Cepsa and UOP, the most modern and efficient technology for the production of linear alkylbenzene (LAB), as well as increasing production capacity at the plant from 200 to 250 Kt/yr

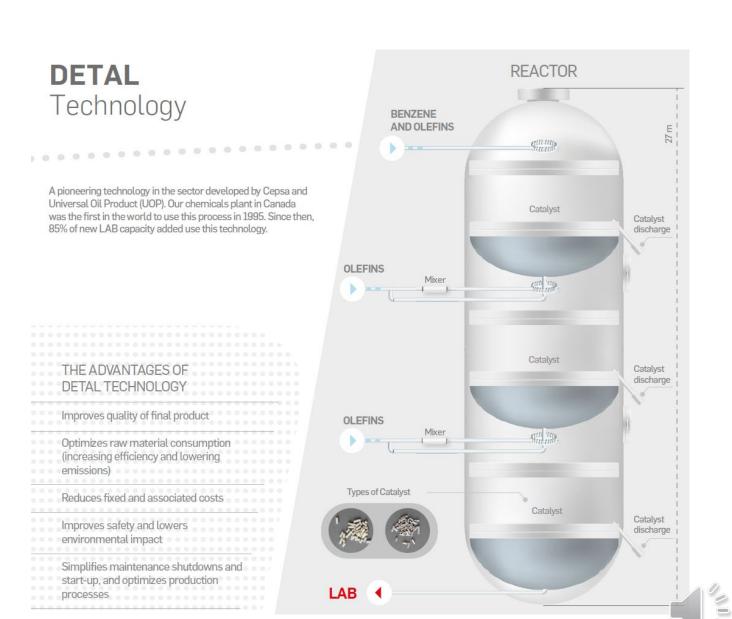


https://www.youtube.com
/watch?app=desktop&v=
uDaLeUb0ij8



DETAL Technology

- The DETAL technology manufactures an improved product, enhance the efficiency of the plant and reduce emissions thanks to lower gas and power consumption and production process optimization.
- The new technology will also significantly improve safety since it is the first in the world that have changed the process from hydrofluoric hydric acid to DETAL technology.
- The new plant has two reactors to alternate production cycles and catalyst regeneration.



YET* 2.0: YET + DYNAMIC SIMULATION



Problems experienced on previously executed YET projects:

- ✓ Phenol 3 (CQP) Fractionation train modeling: lack of data
- ✓ CQPM Paraffins project C₁₃ splitter modeling difficulties: lack of data
- ✓ The only possibility to increase the operating range of the RTO currently is by doing it before at the plant.

What if we do not need the plant running to create our dataset, but instead we create (simulate) them??

✓ Feasibility of predicting the quality of the separation in three specific columns using machine learning models trained with dynamic simulation data.





Inprocess in brief

independent from any ICSS or simulation software provider

our **core business** is Process Simulation

keen to **share its knowledge** with clients



2006

founded in Barcelona by domain experts



years experience



worldwide presence



executed projects



50+

simulation engineers



training courses

Mission: Deliver the value of Process Digital Twins to achieve Operations Excellence



Lifecycle Modelling and OTS projects process

Simulation Lifecycle

Plant start-up support

Plant Lifecycle Conceptual Design, Basic Engineering (FEED) and Detailed Engineering (EPC)

Commissioning &

Operations &

inprocess

Lifecycle Modelling

- · Feasibility studies. Selection of alternatives
- · Dynamic simulation studies before plant construction
- · Validation of control philosophy
- · Operational procedures development/enhancement
- · Process trainer Emulated OTS Early OTS
- · DCS check-out
- OTS for operators' initial and continuous training
- · Support during commissioning and start-up
- · Operations & Maintenance support (Digital Twin)



Process Simulation Studies

- Steady State Analysis
- Dynamic Simulation Studies
- Integrated Flare Systems Analysis
- · Flow Assurance Studies with OLGA
- · Utilities Network Models
- On-line models
- Operations Staff Training
- O&M Support

Simulation Knowledge Profit nprocess

Applications & Software Development

Process Simulation projects

process

Training & Knowledge Transfer

Training / Knowledge Transfer

- · Process simulation courses

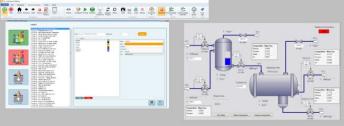




Applications / Software Development

- · IIS: Inprocess Infrastructure Suite
- IPSV: PSVs database
- ITOP: Inprocess Training for OPerators
- ICOM: Inprocess Competence Management System
- IFLOW: to link process simulators with OLGA®
- IPSA: Pressure Swing Adsorption simulator
- OTS Web Access: e-learning options
- Extensions for process simulators







The OTS

- Inprocess was contracted to deliver an Operator Training Simulator (OTS) for Control Room and Field Operators with 3D Virtual Reality
 - Two different ICSS in one OTS
 - Complex 48 steps reactor regeneration sequences
 - Reuse of HYSYS models to develop ML models

Operator Station:

· IIS for Operator Station



Instructor Station and Simulation Server:

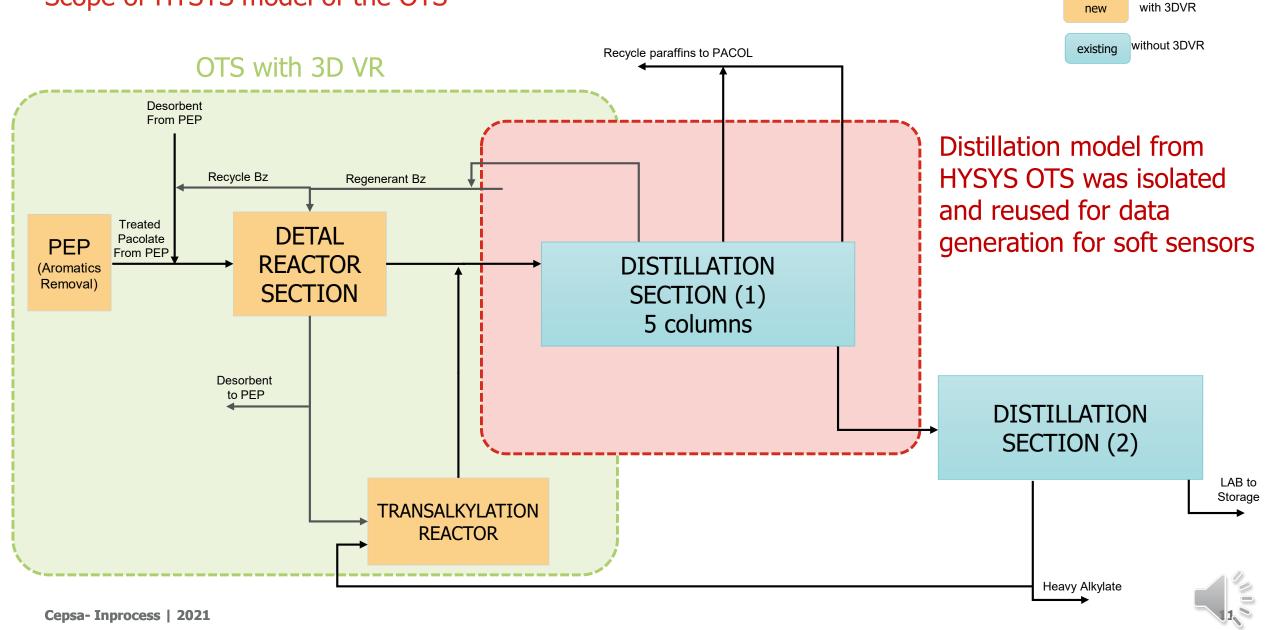
- Inprocess Infrastructure Suite(IIS) for Instructor Station
- Aspen HYSYS® Dynamics Run-time

Virtual Reality Station:

- High-End Nvidia graphic card
- · Virtual Reality Headset

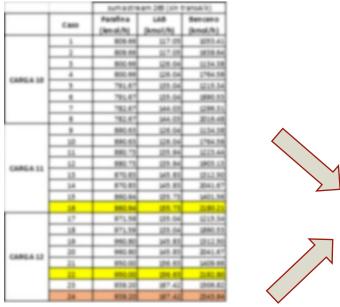


Scope of HYSYS model of the OTS

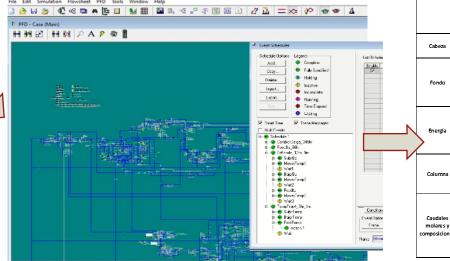


Generation of data for soft sensors from HYSYS Dynamic models Calculated Exported variables

Tested Feeds compositions table



HYSYS Dynamic model of the four columns studied



Caudal

Composic

Caudal molar

Cabeza

Energía

Columna

Composic.

A-V110

caudales modelare

	Caudal L	Stripper Paratinas (A-VSJ)								
	1 [Zona	Variable	Tag	Descripción					
	1 [Caudal	AFRC501	Caudal reflujo					
Reflujo			Caudai	AFRC504	Caudal hacia aspiración A-P10A/B					
	Temperatur			ATRC160	Temperatura reflujo					
	ie mperatur	Reflujo	Temperatura	ATR512	Temperatura A-V53					
	1 1	Kettujo	Temperatura	ATR511	Temperatura salida carcasa A-E53					
	Nivel			ATR513	Temperatura haci a aspiración A-P10A/B					
Cabeza	Te mperatur		Nivel	ALRC502	Nivel A-V53					
	Nivel		Presión	APR503	Presi ón A-V53					
	Presión	Cabeza	Temperatura	ATR526	Temperatura cabeza					
Fondo	Caudal	Capeza	Presión	APRC501	Presión cabeza					
	Caudai	Fondo	Caudal	AFRC503	Caudal fondo					
	Temperatur _	Fondo	Nivel	ALRC501	Nivel fondo					
		Energía	Caudal	AFRC502	Caudal tubos A-ES1					
	Nivel	Energia	Temperatura	ATR507	Temperatura salida tubos A-E51					
	Presión			ı	Columna Renceno (A-)					

	N	vel	ALRC501		Nivel fondo					
	Ca	udal	AFRC502		Caudal tubos A-E51					
	Temperatura .		ATR507	Temps	eratura salida tubos A-E51	1				
	П				Columna Benceno (A					
	1 .		Zona	Variable	Tag	Descripción				
	Temp		Carga	Caudal	AFRC46	Parafinas de bombas de anillo líquido				
3	Temp		Carga	Temperatura	ATR334	TVapores A-V110 a A-V11				
					AFC131	Caudal de reflujo				
	\vdash			Caudal	AFC26	Caudal hacia succión DT-P-04 A/B Temperatura de reflujo				
	Pre		Reflujo	Temperatura	ATR134					
mp	posic.			Nivel	ALRC137	Nivel botellon de reflujo				
			Cabeza	Temperatura	ATR125	Temperatura cabeza				
_	Varia	A-V11	Cabeza	Presión	API834 y APR11	Presión cabeza				
	Cau			Caudal	DTFI341	Caudal fondo				
	Niv		F 4		AT1870	Temperatura fondo				
	Tempe		Fondo	Temperatura	ATRC509	Temperatura hacia A-V 110				
	Pres			Nivel	ALRC118	Presión en el nivel de fondo				
	Cau				ATR123					
	Tempe		Columna	Temperatura	ATRC124	Temperaturas columna				
_	Niv				ATR120					
	Cau			Caudal	AFRC140	Caudal fondo				
	Tempe									

ALRC330

AFRC122

AFRC327

ATR121

ATR329

ATRC509

ATR508

ATR333

ATR332

INF_AV11_PAR_TOP

462_Parafinas

Tested Input variables tables

						D	T-V-03												1
Zona		Variable	5	Та		Valor Sacyr	Valor Train1	Valor Train2	Valo Train			Rango Min	Rango Max	Salto					
		ga PACOL (: d) (corrienti										10	12	1					,
	0	le/Para (mol:	mol)									10	13	1					
Carga	(co	2/5 /	4.00				C	olBz (A	/-11/	A-V110)	10			_				į
	Rat (i	Zona		Va	ariables	s		Tag	1	Valor Sacyr	ur	idades	Rang Min	o Ran Max		Salt	0		(
	Marc	Canan	asos de carç	ga: variación LAB/Paraf] `		
	Temp	Carga								AV-5	MI	NY14							
Cabeza		Cabeza		-		Variables		\neg	Tag		Valo	r	idades		tango Rango		enden.		
- ,		Energía	Bzfa	Zona			arnatike	•		149	_	Sacy		Garden.	M	n	Max	Salto	1
Energía		_	\vdash	Carga		3.0	area or	ga											
		Presion	\perp	Cabeza	Refujo A-VSI					APRCS11.5P		175.4		m3/id					3
					Refuge A-Y34				=	APRC157.5P		1060		m3/d					ä
						Preside AVIA (mm/q)			_	APRC15	<u>1.56</u>	-764	_		_	481	-750	20 sp	Ц
					Jame et	98525.:		refress.A		ATRCSS	<u>1,5P</u>	232.4		ng C	_				Э
				Energia	Energia Tempylate34 -> Nabil 636			com para	un paral. A-		ATR C3 46-5P		0	leg C					1

Models were split in several cores running for 3-4 days at 15 Real time factor to represent about 1600 different plant status.

CEPSA was also taught to run their own runs.



Nivel fondo

Caudal tubos A-E15

Caudal tubos 4-F150

Temperatura salida tubos A-E15

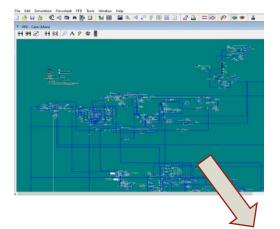
Temperatura salida tubos A-E150

Temperatura salida carcasa A-E52

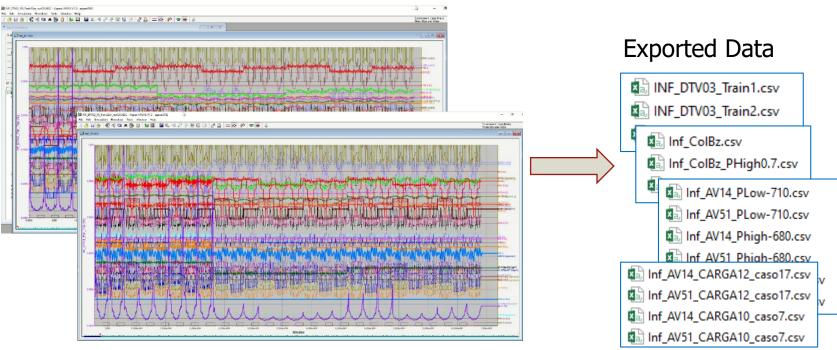
Temperaturas columna

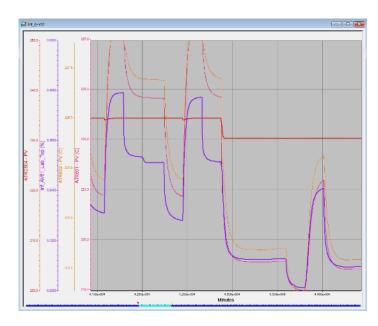
Caudal de parafinas en corriente 462 (fond

Virtual process data generated by HYSYS Dynamics tests



The data contained key compositions (impurities in top or bottom of distillation columns) which are not directly measured in the plant

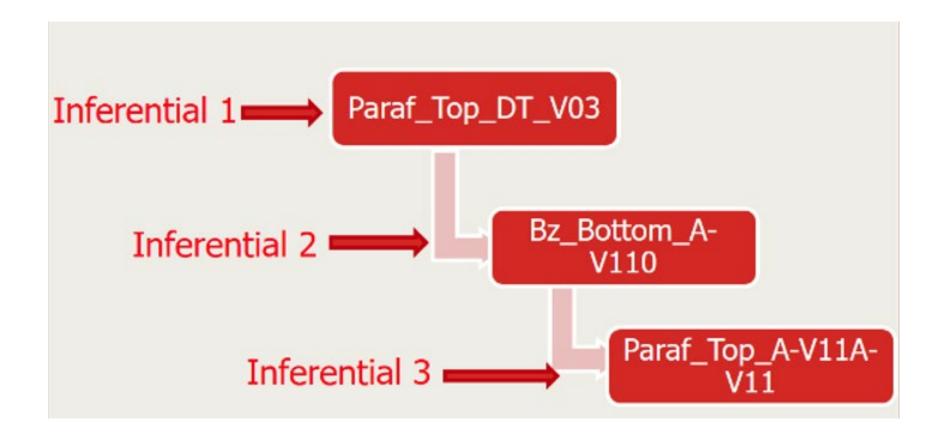




CEPSA used their Data Scientists to train their Machine Learning models and deploy them in the plant



Machine learning models

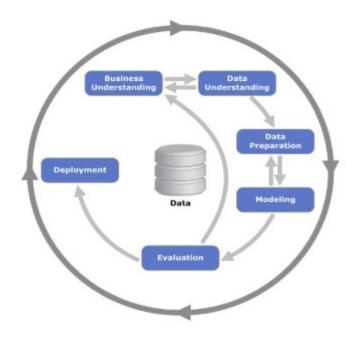




Machine Learning modeling methodology

 Machine Learning is a branch of the Artificial Intelligence whose goal is to build systems that automatically learn from data and adjust actions or decission accordingly.

Cross-industry standard process for data mining (CRISP-DM)



Software



Python was used to do the exploratory analysis, data preparation and models of this project.

Advantages:

- Open source.
- Easy to use.
- Flexible and fast to perform flexible exploratory analysis.
- Easy to find internet errors helps

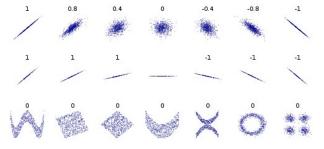
Disadvantages:

· The need of programing skills

Note: Aspen Hybrid Models were not available at the time of the project

Data discovery and analytics

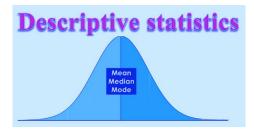
Correlated variables



The workflow to determine variables correlation was:

- 1- Set group of correlated variables
- 2- Select the most important variable with respect to the objective and eliminate the rest of them

Descriptive Statistic



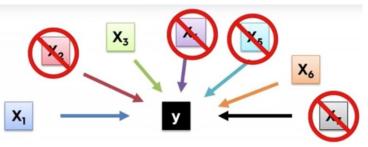
Summarizing and organizing the data so they can be easily understood



The feature selection is the process that choose a reduced number of explanatory variable to describe a response variable.

- 1- Make the model easy to interpret
- 2- Reduce the size of the problem allowing algorithms to work faster
- 3- Reduce overfiting

- √ Lasso
- √ Gradient Boosting
- √ Random Forest
- √ Recursive Hybrid elimination*
- * Several times iterated



Variables group

The groups were built based on the importance of the variables and the knowlege of the business



Data preparation

Bz Bottom A-V110:

- 33 Variables (columns) generated by Hysys
- 85,848 rows
- All numerical variables
- Determination of minimum, maximum, average and standard deviation values, distribution histograms.

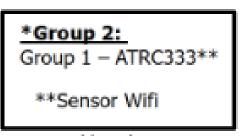
Correlated variables:

9 groups of correlated variables were determined. <u>Determination of the most important variables:</u>

- ✓ Lasso
- ✓ Gradient Boosting
- ✓ Random Forest
- ✓ Recursive Hybrid elimination*
 - * Several times iterated

Based on the observed before and knowledge of business, the following groups of variables were determined:

Group 1	
API0834	
ATR333	
ATRC332	
ATR134	



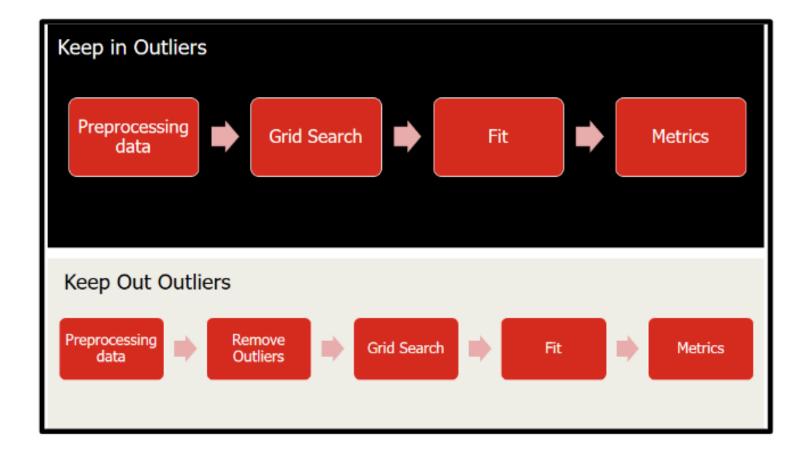
Group 3
API0834-ATR134/ATRC332-ATR134
ATRC332
ATRC332-ATR134/API0834

Group 4
API0834
API0834-ATR134/ATRC332-ATR134
ATRC332
ATR134

The groups were built based on the correlations of the variables with the target and process knowledge.

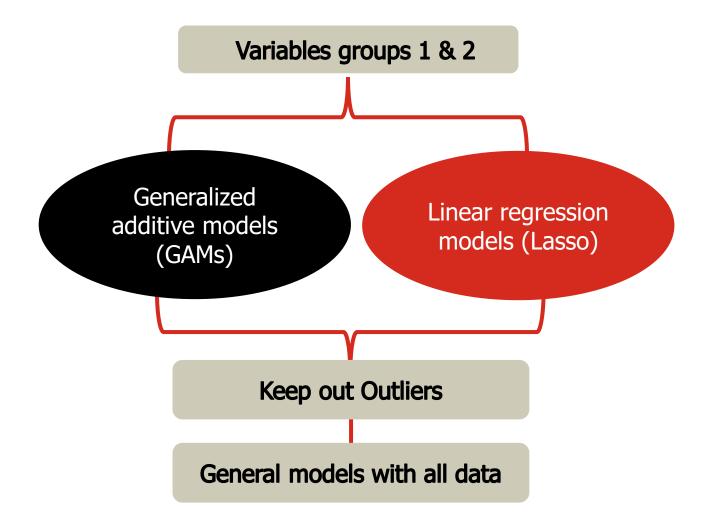


Machine learning modeling



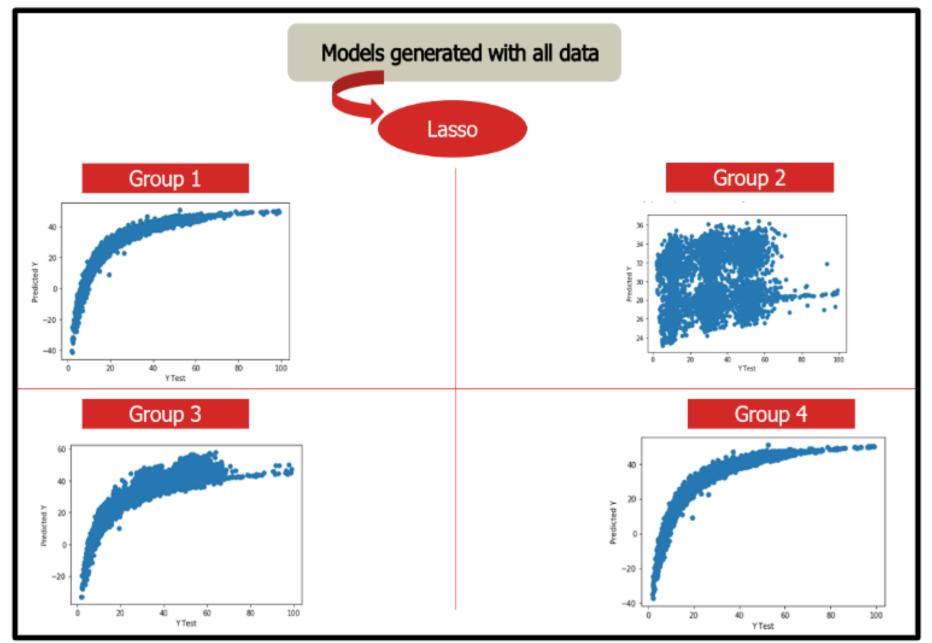


Machine learning modeling



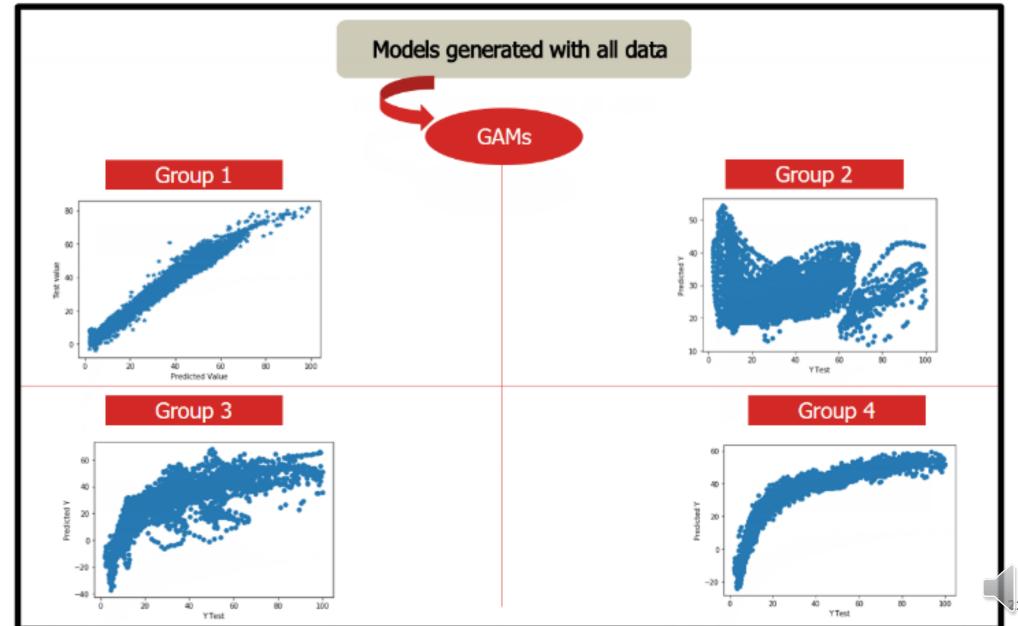


Results obtained from Lasso Models

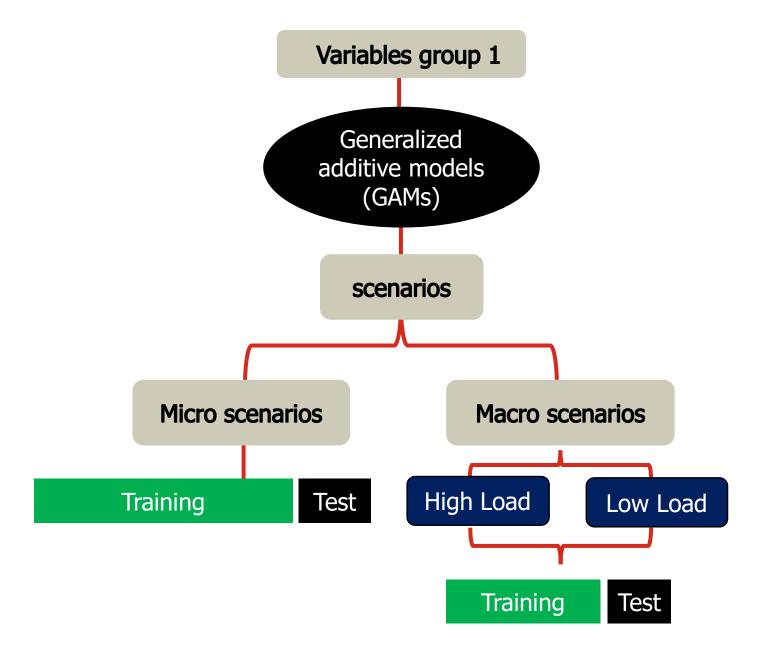




Better results obtained from GAM Models

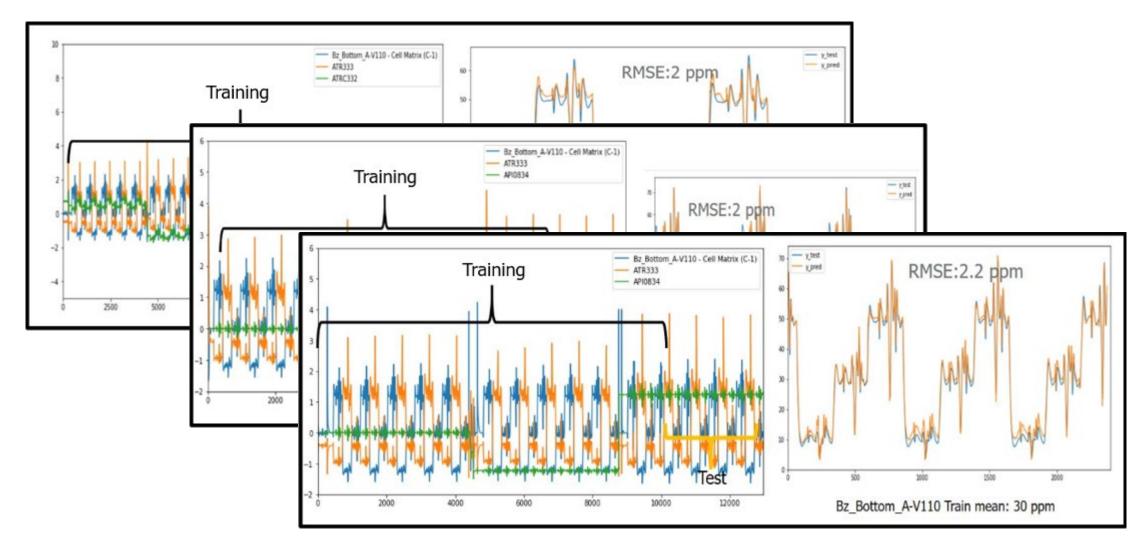


Model Testing





Model Testing





Conclusion

- 1.5% energy saving on each column involved.
- Quality and recovery improve on the top and bottom streams of the columns.
- Payback around 1 year or less. Minimum investment and good returns.
- Technically viable the data extraction from simulation to create and train machine learning models, even before plants start-up.





Thank you

