

Sustainable design of reverse osmosis membrane processes in ingredient production

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The Food & Beverage industry faces multiple challenges ...

External

INFLATION



SUPPLY CHAIN DISRUPTION



SUSTAINABILITY



CONSUMER DEMANDS



Internal

SPEED OF INNOVATION



WORKER SHORTAGES



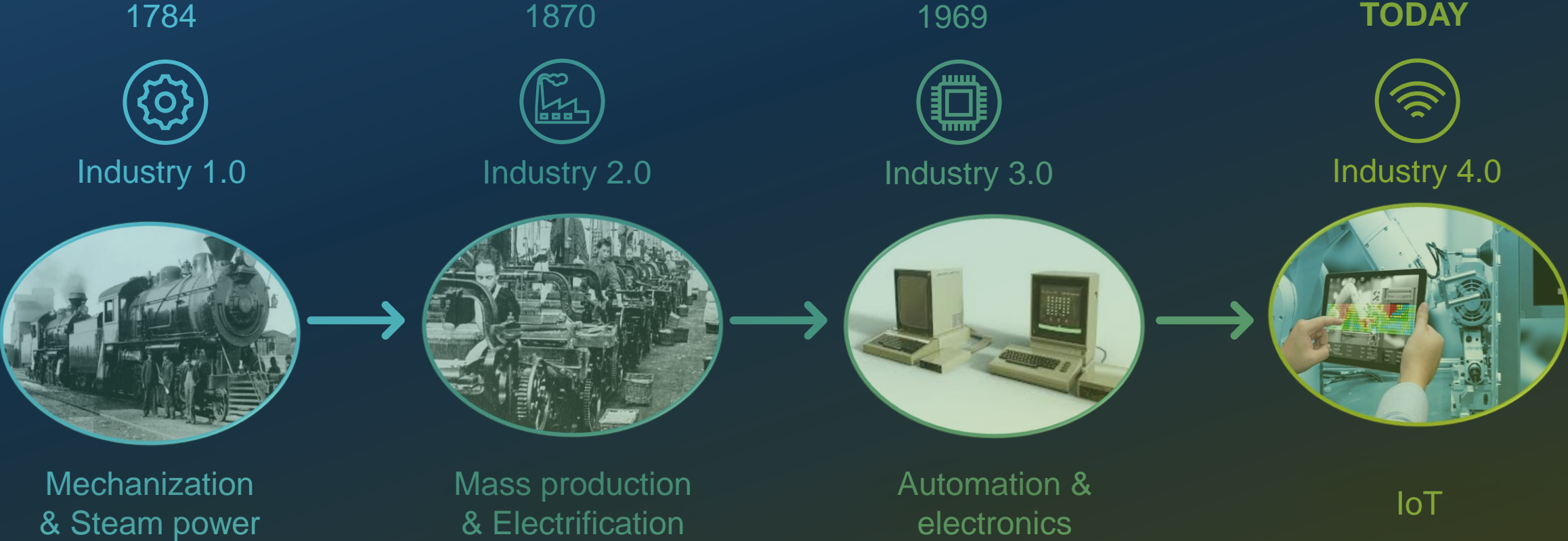
DATA TRANSPARENCY



GLOBAL STANDARDIZATION



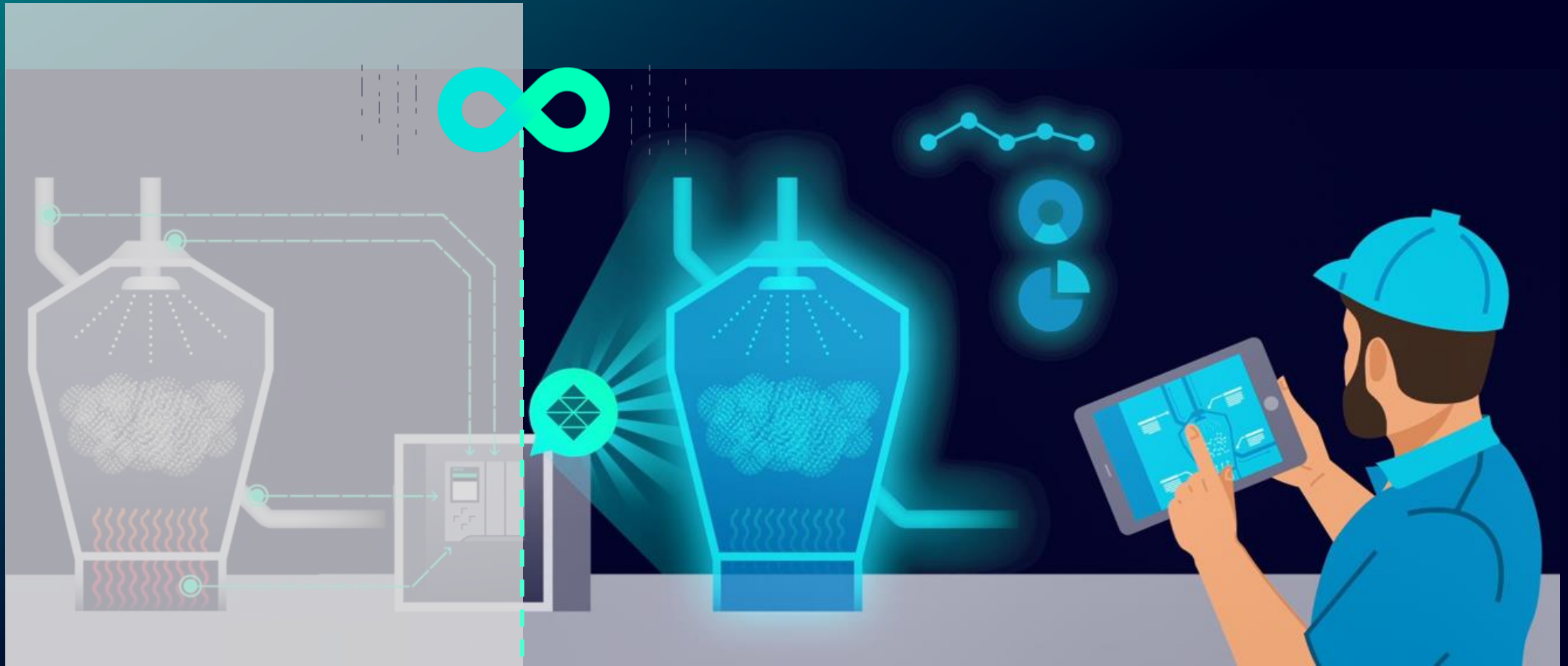
The F&B industry is evolving again to meet these challenges



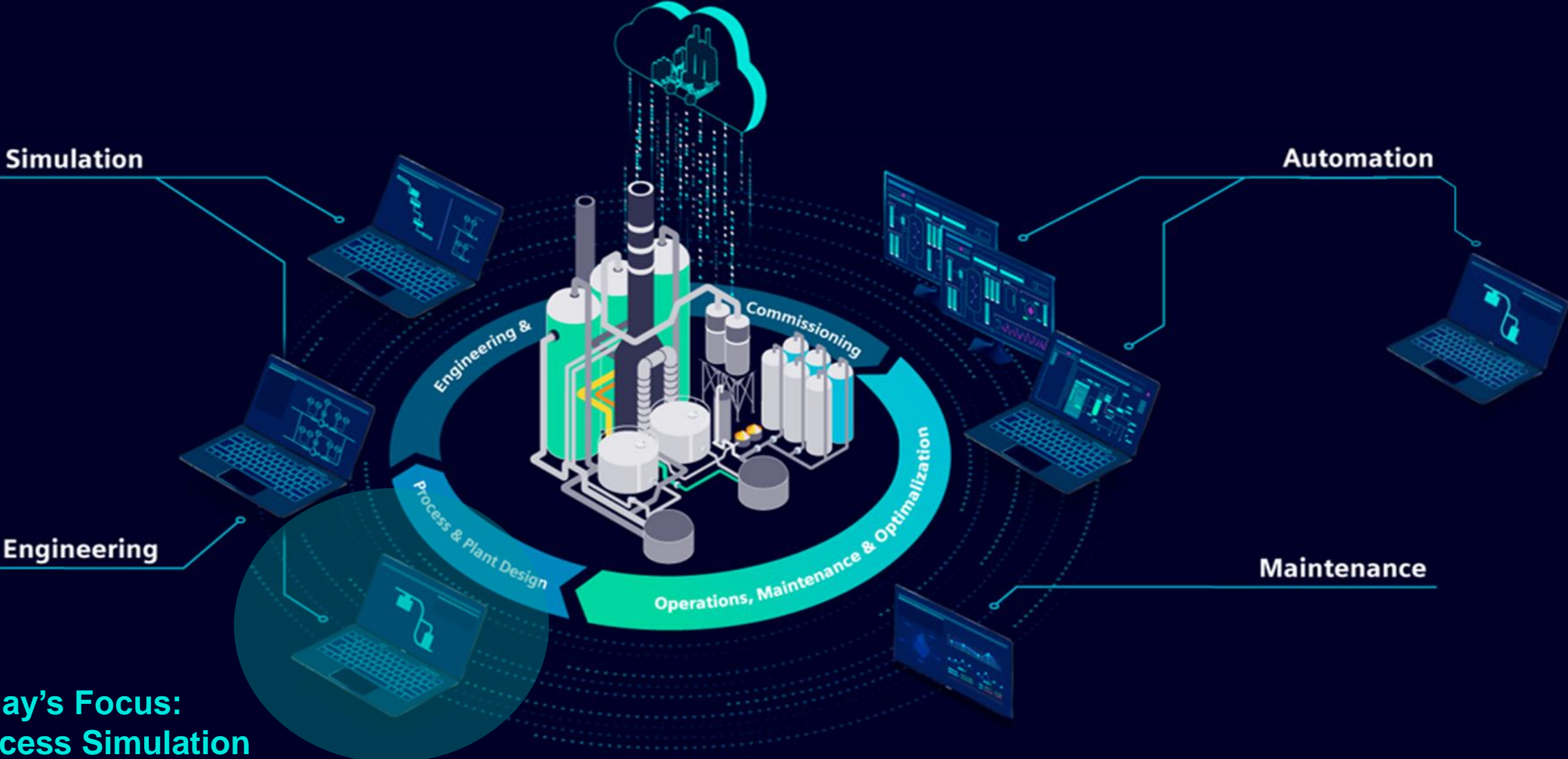
Process automation & optimisation is one of the biggest focus areas for innovation by food manufacturers, **73% of manufacturers** are investing in process automation & optimisation

A new path for the food and beverage industry

Status quo is insufficient to achieve production excellence, **digital transformation** is required.



Digital Twins – Not only process models






**Today's Focus:
Process Simulation**

Overcome innovation barriers to sustainable F&B production with digital process twins

Current situation:

- Cost pressure means that over **65%** of companies are prioritizing product innovation and process efficiency.
- Corporate responsibility targets drive energy, water & waste reduction.
- Consumer quality demands are increasing.
- **Result:** F&B companies risk needing to invest **€ millions** to stay competitive

Typical benefits:

-  Increase process line uptime by up to **10%** to untap **€3-5 M** annual revenue
-  Avoid **€15-20 M** CAPEX spend in multi-factory process upgrades
-  Reduce water and evaporation energy demand by **>50%** to remove impurities

What questions might be asked to increase your company's speed of innovation and competitiveness?



How can I increase run time of a protein concentration process to meet market demand?



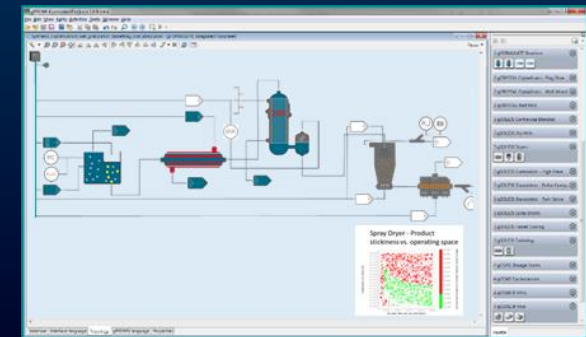
How can I design a safe process for multiple products with no more than necessary CAPEX spend?



How can I design an impurity removal process with minimal water & energy usage?



How can I produce many high quality recipes with a shorter batch time to differentiate my product in the market?



The solution: Digital design of Food & Beverage processes using physics-based process digital twins

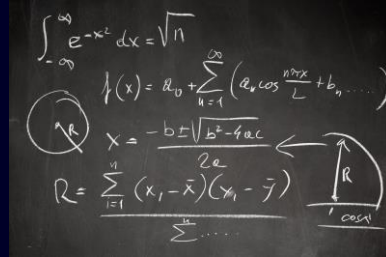
Advantages of physics-based process modelling

- Two types of modelling:

Data-driven

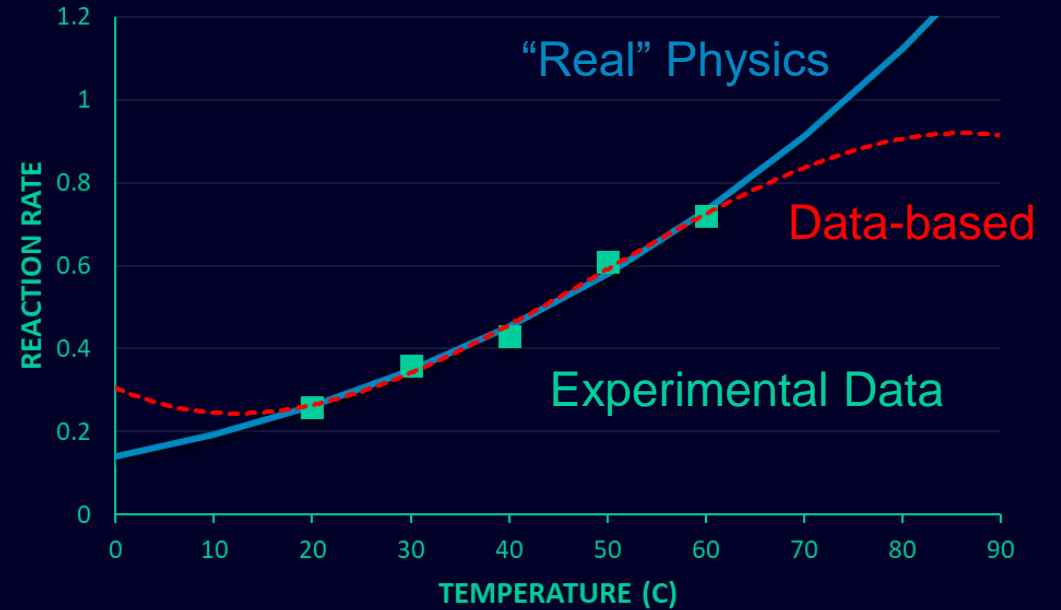


Physics-based



- Physics-based modelling:
 - Requires less experimental data
 - Can be used outside of data range
 - Transfers better to other scales/equipment
 - Takes advantage of well-known phenomena

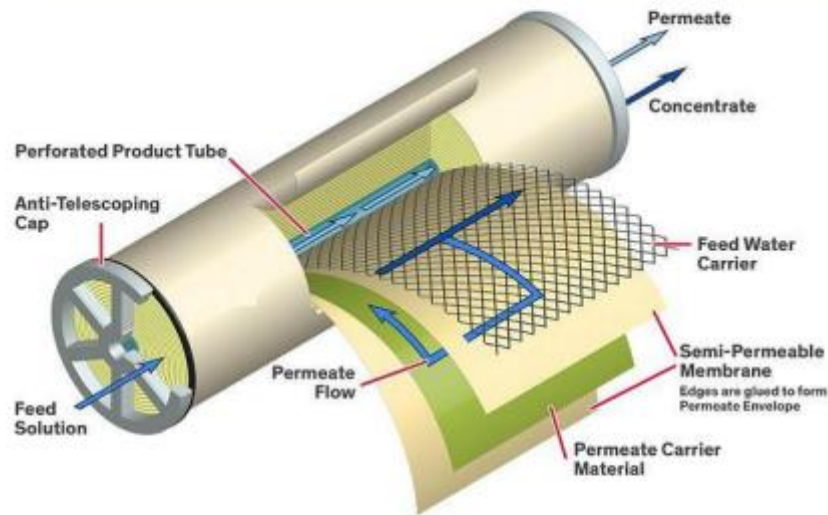
Model Comparison





Membrane System Modelling

TANGENTIAL FLOW MEMBRANE



Membrane flux

$$J_p = \frac{P_{TMP}}{R}$$

⊖ Notation

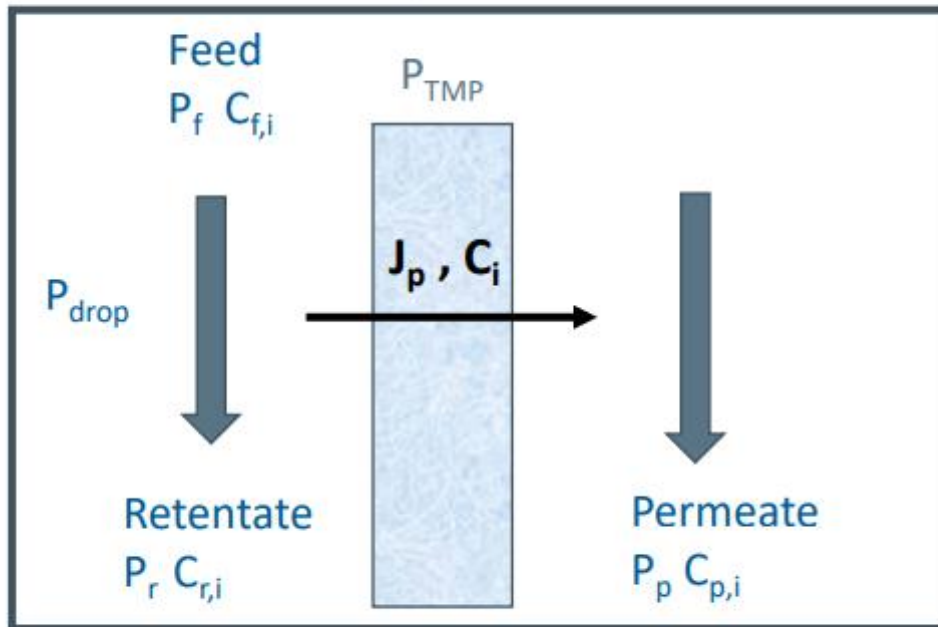
J_p	Membrane flux to permeate	$\text{kg m}^{-2} \text{s}^{-1}$
P_{TMP}	Transmembrane pressure	Pa
R	Membrane resistance, $R = R_{dyn} + R_{stat}$	$\text{Pa m}^2 \text{s}^1 \text{kg}^{-1}$

Membrane pressures

$$P_{TMP} = P_f - \frac{P_{drop}}{2} - P_p$$

⊖ Notation

P_{TMP}	Transmembrane pressure	Pa
P_f	Feed stream pressure	Pa
P_p	Permeate pressure	Pa
P_{drop}	Pressure drop	Pa



Rejection coefficient

$$f_{rej,i} = 1 - \frac{C_{p,i}}{C_{f,i}} \quad i \in \text{active components} - \text{water}$$

⊕ Notation

$f_{rej,i}$	Rejection coefficient	-
$C_{p,i}$	Mass concentration of component i in permeate stream	kg/m^3
$C_{f,i}$	Mass concentration of component i in feed stream	kg/m^3

FLUX MODELING

J_p Permeate flux $\left[\frac{kg}{m^2 \cdot s}\right]$

P_{TMP} Transmembrane pressure $[Pa]$

R Membrane resistance $\left[\frac{m^2 \cdot Pa \cdot s}{kg}\right]$

$$J_p = \frac{P_{TMP}}{R} = L_p \cdot (P_{TMP} - \Delta\pi_{real})$$

$$R = R_{water} \cdot R_{product}$$

$$R_{water} = \frac{1}{L_p}$$

$$R_{product} = \left(1 - \sigma \cdot \left(\frac{\Delta\pi_m}{P_{TMP}}\right)^p\right)^n$$

$$J_p = \frac{P_{TMP}}{\frac{\left(1 - \sigma \cdot \left(\frac{\Delta\pi_m}{P_{TMP}}\right)^p\right)^n}{L_p}}$$

L_p water membrane permeability $\left[\frac{kg}{m^2 \cdot Pa \cdot s}\right]$

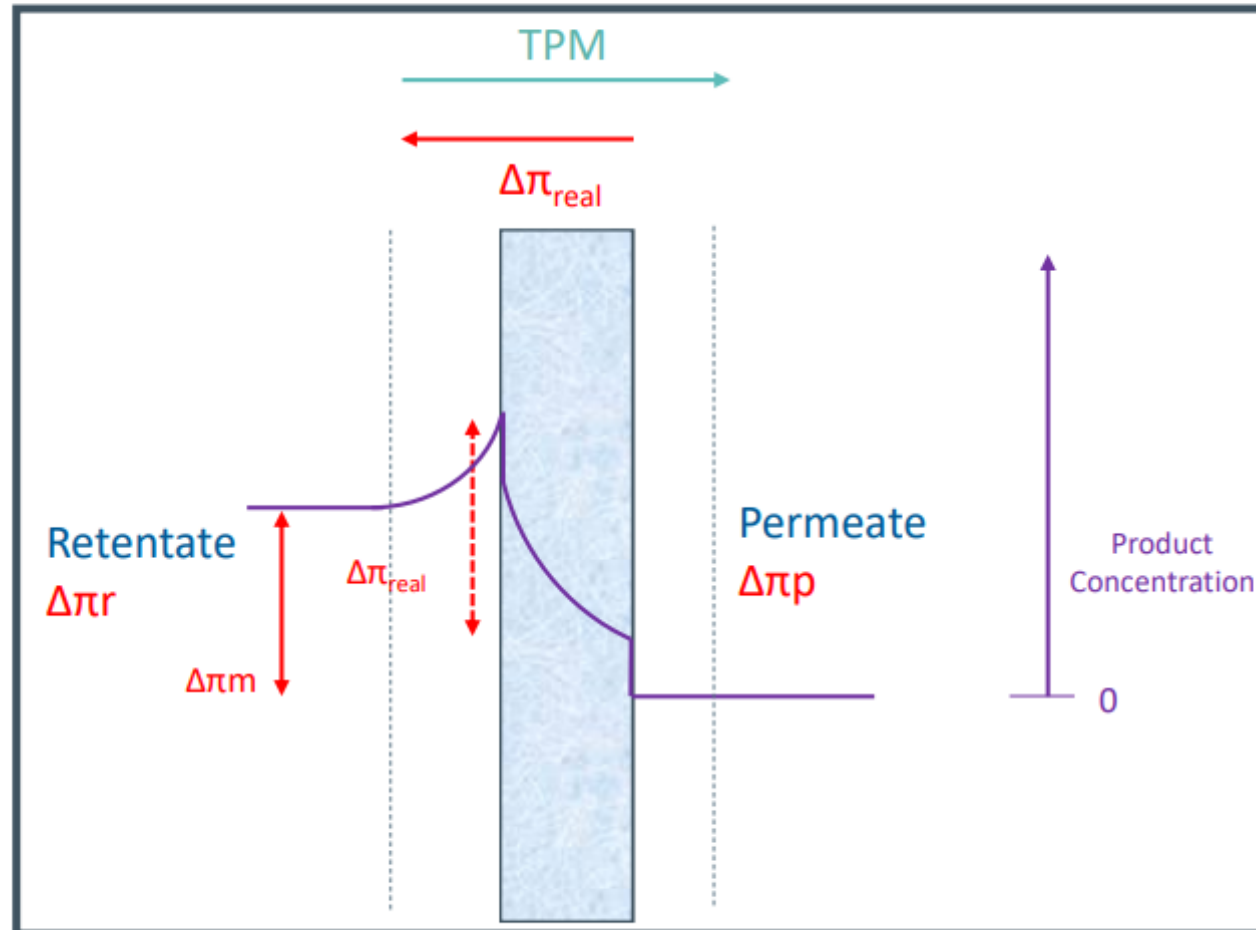
R_{water} water membrane resistance $\left[\frac{m^2 \cdot Pa \cdot s}{kg}\right]$

$R_{product}$ product membrane resistance

$\Delta\pi_m$: osmotic pressure membrane difference $[Pa]$

σ reflexion coefficient

n & p reflexion coefficient parameters



FLUX MODELING

Water activity & Osmotic pressure

$$\Delta\pi = -\left(\frac{RT}{V_m}\right) \cdot \ln a_w$$

$\Delta\pi$ osmotic pressure [Pa]

$$R = 8314 \left[\frac{l \cdot Pa}{mol \cdot K} \right]$$

V_m molar volume (water) $18,07 \cdot 10^{-3} \left[\frac{l}{mol} \right]$
 T [°K]

$$\ln(a_w) = \frac{-\xi m}{55.1}$$

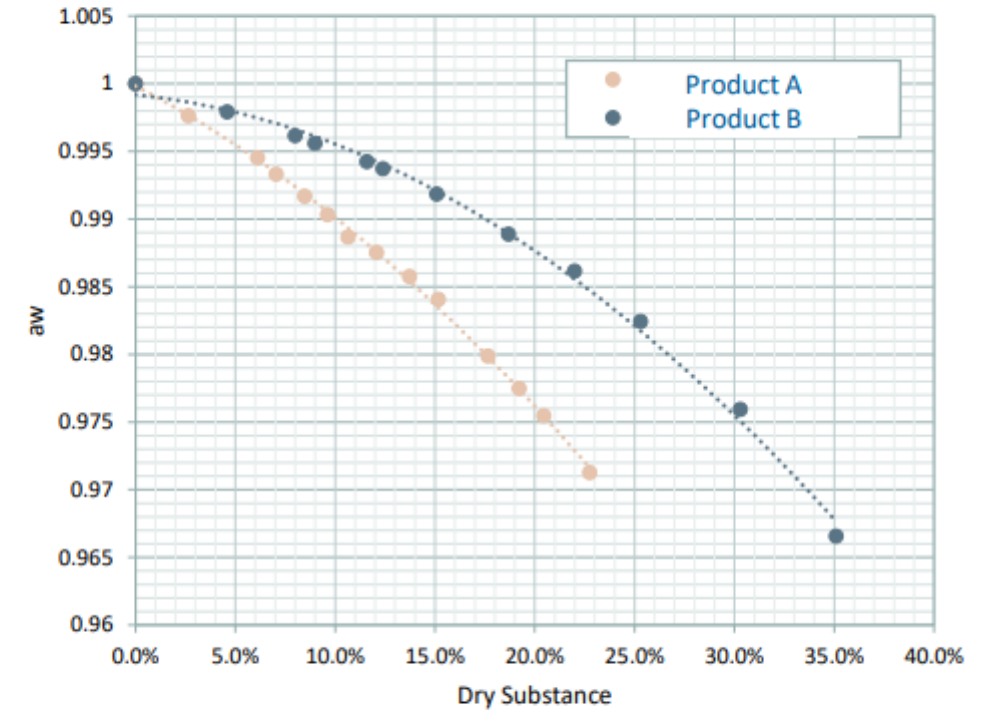
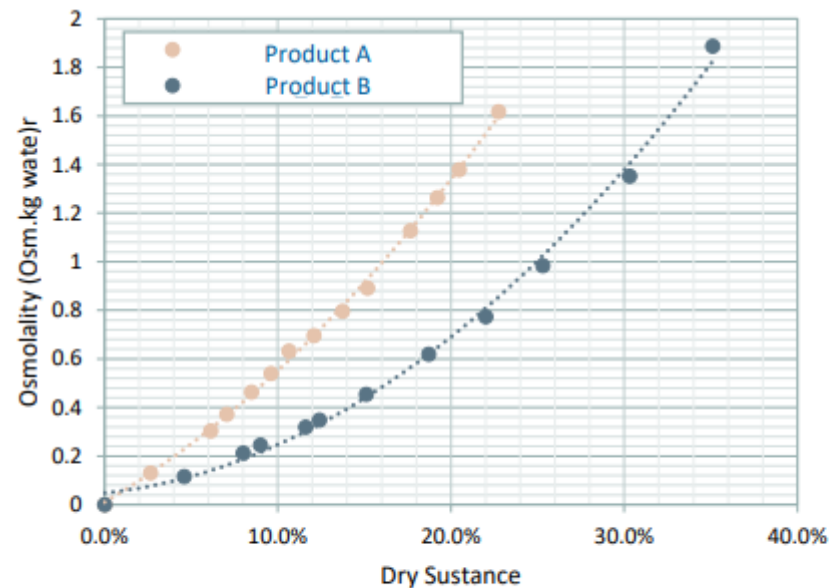
$$a_w = e^{\left(\frac{-\xi m}{55.1}\right)}$$

a_w water activity
 ξm Osmolality $\left[\frac{Osm}{kg \text{ water}} \right]$



OSMOMAT 3000

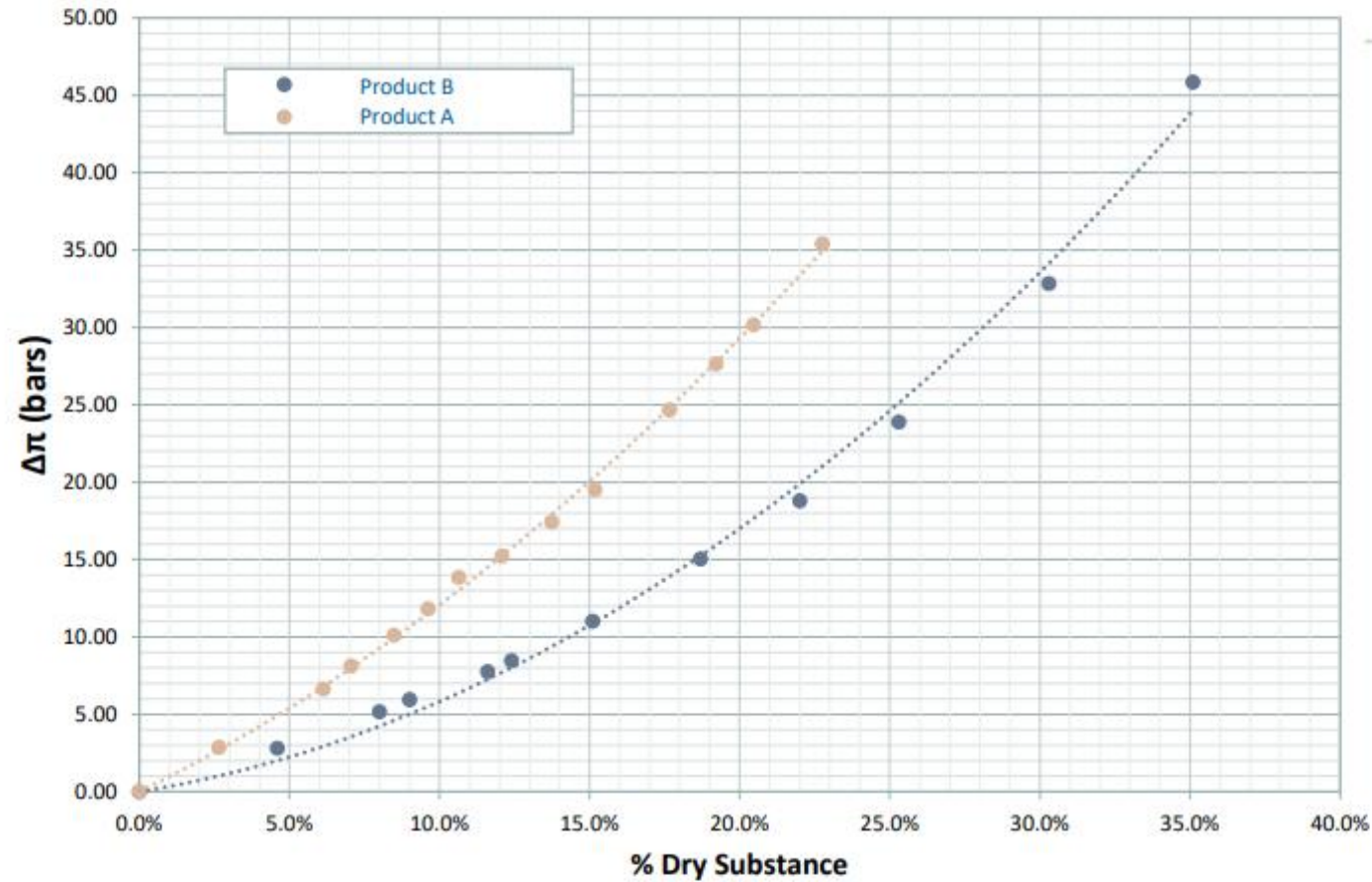
Osmolality ξm (mOsm/kg water)



$$a_w = a_{w2} \cdot DS^2 + a_{w1} \cdot DS + a_{w0}$$

Coefficient	PRODUCT B	PRODUCT A
a_{w2}	-2.134E-01	-2.106E-01
a_{w1}	-1.496E-02	-7.615E-02
a_{w0}	1	1

OSMOTIC PRESSURE $\Delta\pi$

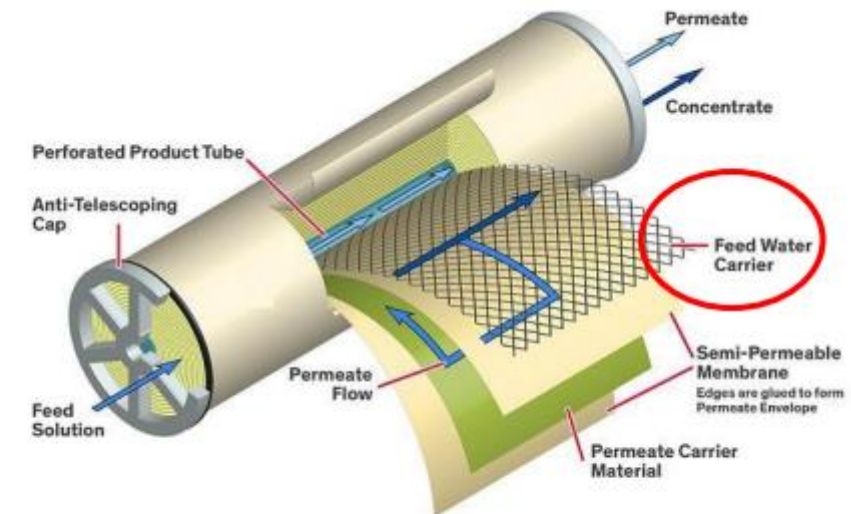


Designation	Characteristics	NaCl rejection
RO90	Thinfil composite	$\geq 90\%^{**}$

** measured on 2000 ppm NaCl, 9 bar, 25°C

Part No.	Element type	Area	
		m ²	(sqft)
527977	RO90-8038/30	38	(409)
525510	RO90-8038/18	28.1	(302)

Lp
(kg/m²/Pa/s)
1.20E-08
1.40E-08





Overview Case Studies

Whey protein concentrate (WPC) production – Improve uptime of ultrafiltration process to meet demand





Customer challenge

- To reduce protein fouling in a WPC ultrafiltration process to increase process uptime & reduce CIP frequency
- To meet product specification and capacity requirements

Solution

- Configure and calibrate **physical science-based process model** of ultrafiltration process
- Use optimisation techniques to identify operating conditions which reduce rate of fouling and increase run time

Customer benefits

-  Reduced membrane fouling delivering up to **10% increase** in process runtime
-  Higher productivity and reduced CIP
-  Streamlined experimentation and process upscaling to engineering and operations
-  CAPEX savings



Creating value with APM: WPC

Unlocking the APM potential

- Organization**
 - Organisation of the technology chain
 - Managed team based development program
 - Alignment to supply chain: timely demand driven development of models and libraries (KISS)
 - Internal PR to/through senior management
- People**
 - Multi-tier modelling knowledge in the chain through training program → PSE can help here
- Tooling**
 - User friendly, complete, stable and documented modelling tools. Multi-tier user interfaces.

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Recovery of beer from yeast slurry – Improve process understanding and consistency

Customer challenge

- Yeast slurry after beer fermentation contains a significant amount of good beer
- Equipment exists to separate beer from yeast, but process **know-how** is lacking

Solution

- Apply membrane models to calibrate the process and develop better process understanding
- Deploy models to aid in process prediction and operational decision-making

Customer benefits



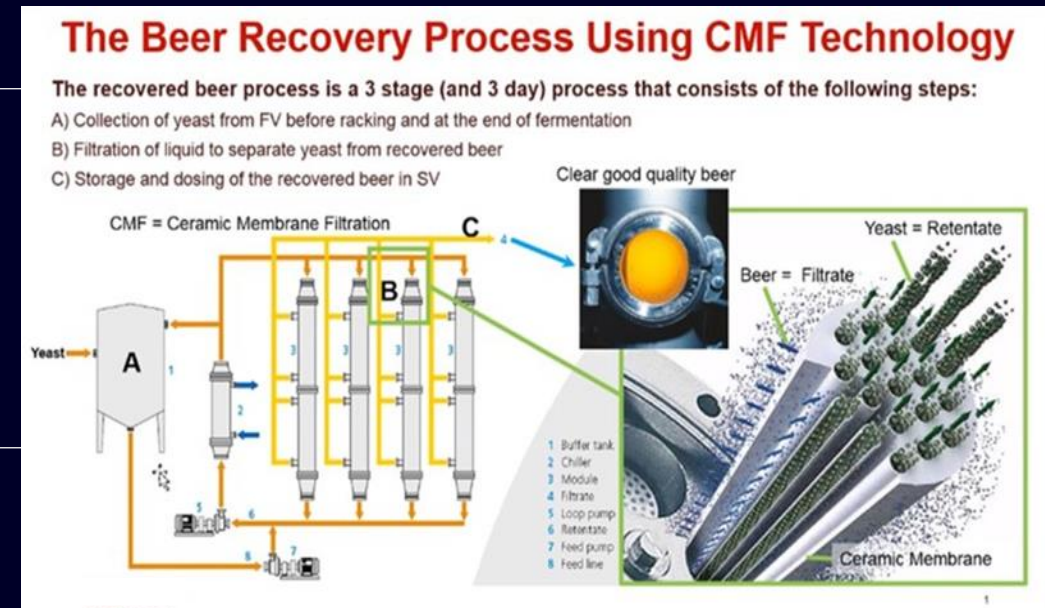
Reduce operator efforts in process control



Improve beer removal from 60% to 80%



Financial benefit of \$50-200k per annum, depending on brewery size



Plant-based ingredient production – Sustainable design of reverse osmosis impurity removal process

Customer challenge

- **Sustainable** process design – performance evaluation of different configurations
- **Minimize energy and water consumption** in process
- Improve quality of product given varying feed impurity levels

Solution

- Configure and calibrate **physical science-based process model** of the reverse osmosis process in **SIEMENS gPROMS FormulatedProducts**
- Performance analysis using global system analysis
- Use optimisation techniques in gPROMS to identify optimum operating conditions

Customer benefits



Better, leaner and more competitive designs



20% reduction in diafiltration water consumption



65% reduction in evaporation energy

Operating space discovery for optimal and consistent product quality





Reverse Osmosis Process Design

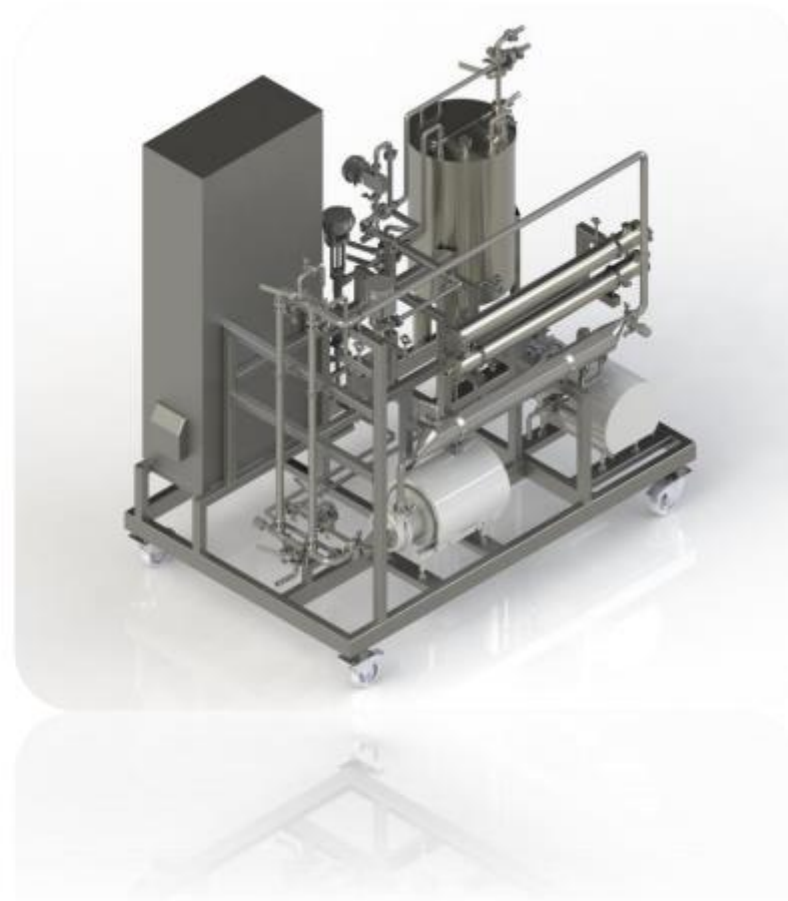
SCOPE OF STUDY

Quality improvement of our products

Remove impurity with a new membrane process

Development of modeling tool

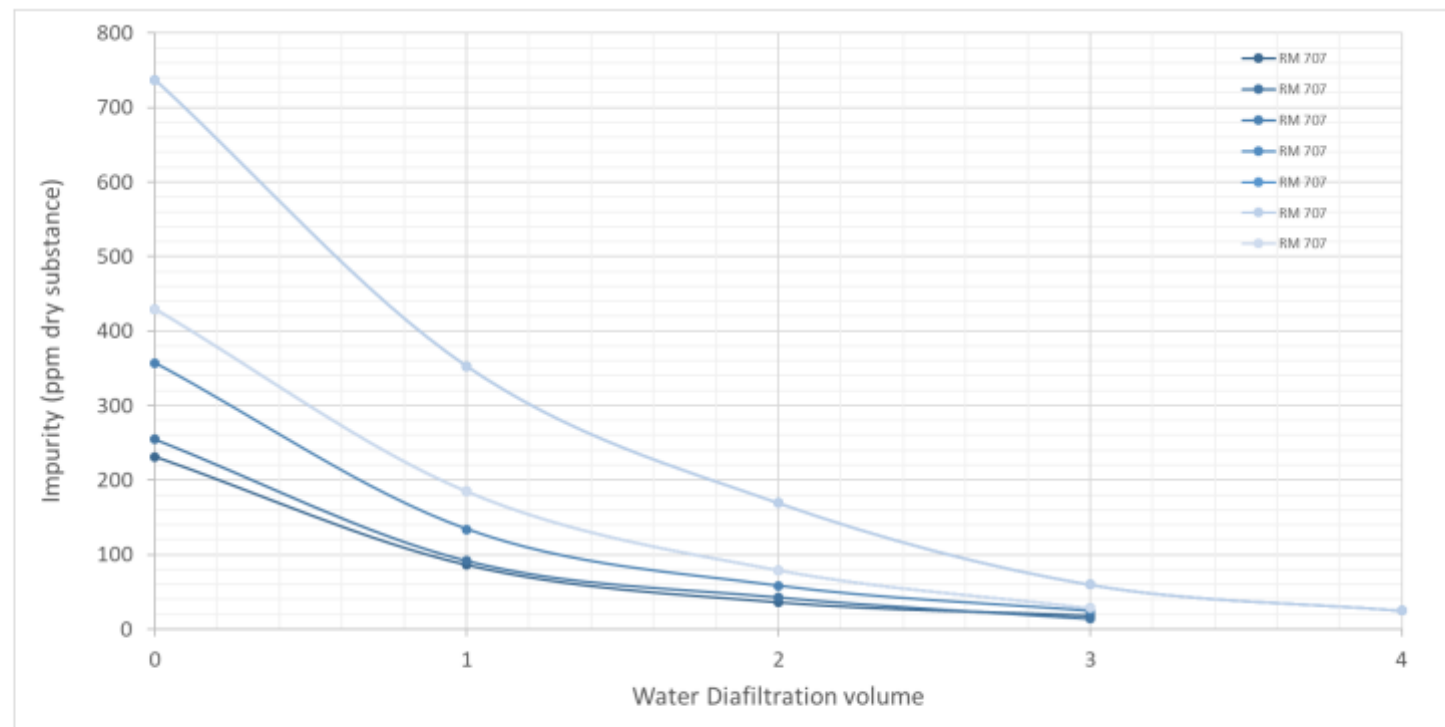
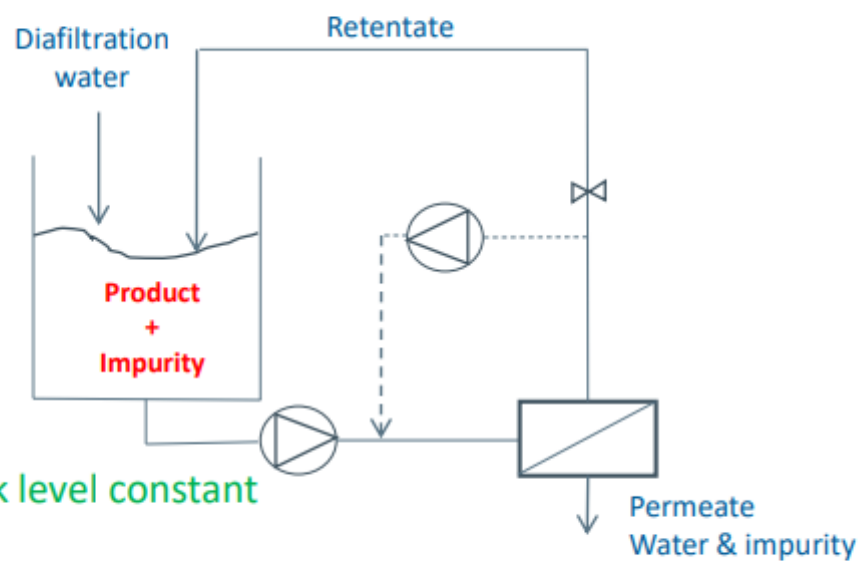
- New model for NF/RO
- Batch system
- Experiment at pilot scale



Tool for process performance evaluation and scale-up

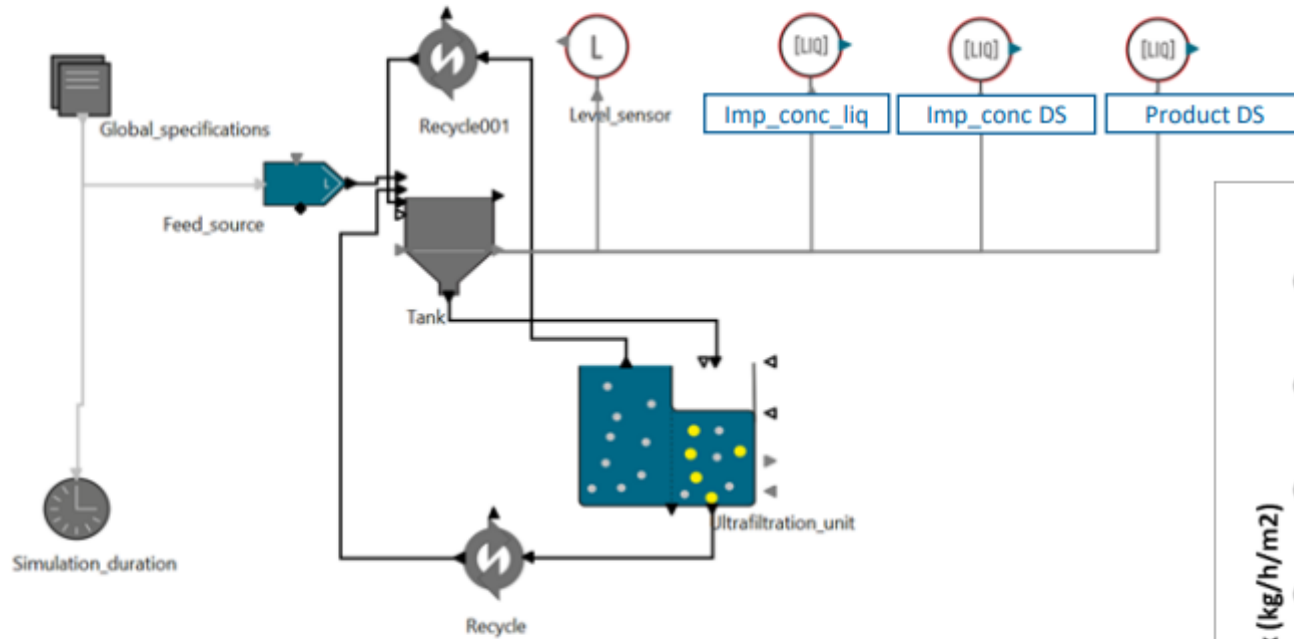


PROCESS PILOT SCALE



Scale-up to continuous ?

FLUX MODELING – MODEL VALIDATION



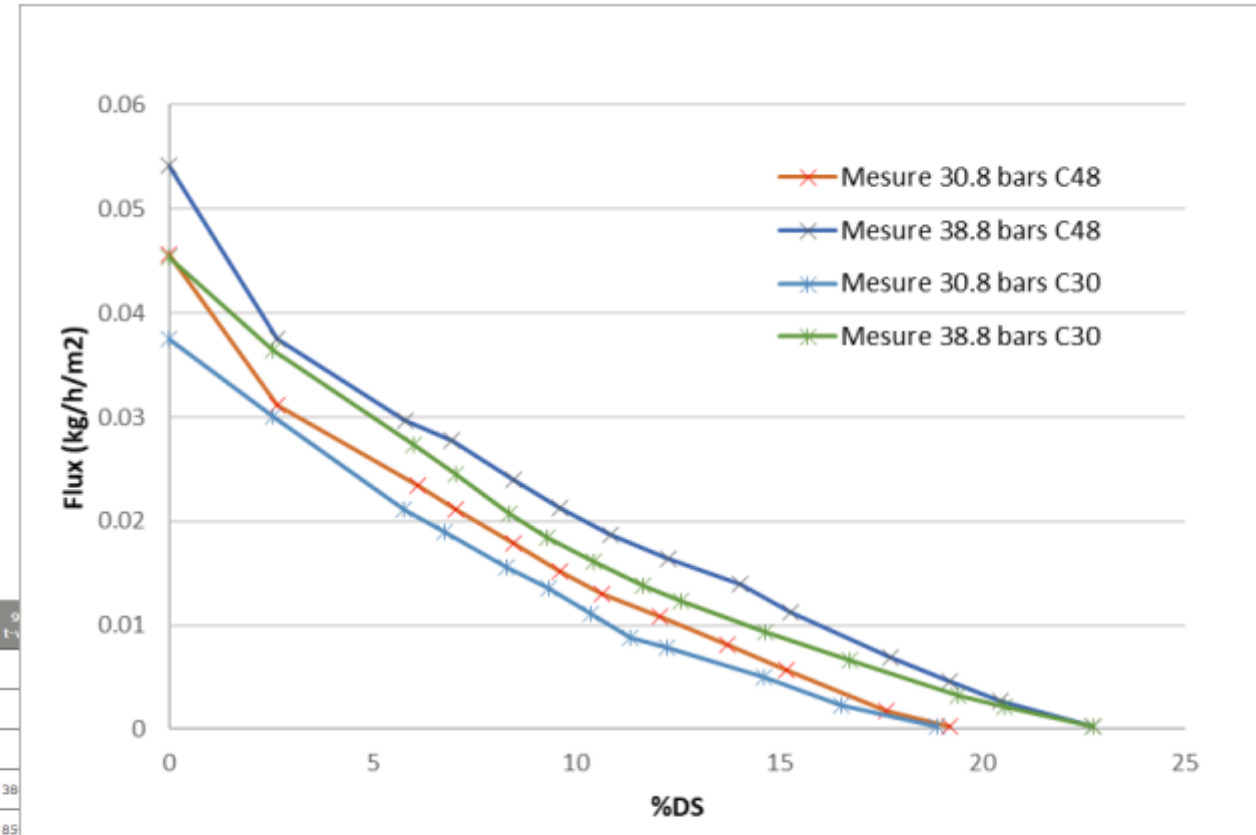
Model Parameters

- Probability of parameter lying between (Final Value -4% Confidence Interval) and (Final Value +4% Confidence Interval) = 4%
- The t-value shows the percentage accuracy of the estimated parameters, with respect to the 95% confidence intervals

Model Parameter	Final Value	Initial Guess	Lower Bound	Upper Bound	Confidence Interval			t-value
					90%	95%	99%	
Ultrafiltration_unit Static resistance: Custom empirical model - Custom resistance parameters (aw0)	1.00000	1.00000	1.00000*	1.00000*				
Ultrafiltration_unit Static resistance: Custom empirical model - Custom resistance parameters (aw1)	-0.0149600	-0.0149600	-0.0149600*	-0.0149600*				
Ultrafiltration_unit Static resistance: Custom empirical model - Custom resistance parameters (aw2)	-0.213360	-0.213360	-0.213360*	-0.213360*				
Ultrafiltration_unit Static resistance: Custom empirical model - Custom resistance parameters (n)	-1.61506	-1.61506	-2.00000	-1.00000	0.00349131	0.00418167	0.00556622	38
Ultrafiltration_unit Static resistance: Custom empirical model - Custom resistance parameters (sigma_cst)	0.835206	0.835206	0.00000	1.00000	0.000811576	0.000972054	0.00129390	85
Ultrafiltration_unit Static resistance: Custom empirical model - Custom resistance parameters (sigma_power)	0.520874	0.520874	0.00000	1.00000	0.000490829	0.000587884	0.000782532	886.016
Reference t-value (95%):								1.67252

[Click here](#) to use above final values in future calculations

* a parameter that lies at one of its bounds is excluded from the statistical analysis



FLUX MODELING

$$J_p = L_p \cdot \frac{P_{TMP}}{\left(1 - \sigma \cdot \left(\frac{\Delta\pi_m}{P_{TMP}}\right)^p\right)^n}$$

Model	Model Osmolality/falaval RO90-48		Model Osmolality/falaval RO90-30	
Lp (kg/m²/Pa/s)		1.40E-08		1.20E-08
Produit	PRODUCT B	PRODUCT A	PRODUCT B	PRODUCT A
σ	0.835	0.924	0.835	0.924
p	0.521	0.566	0.524	0.571
n	-1.61	-1.20	-1.61	-1.20
MAPE(%)	8.5			

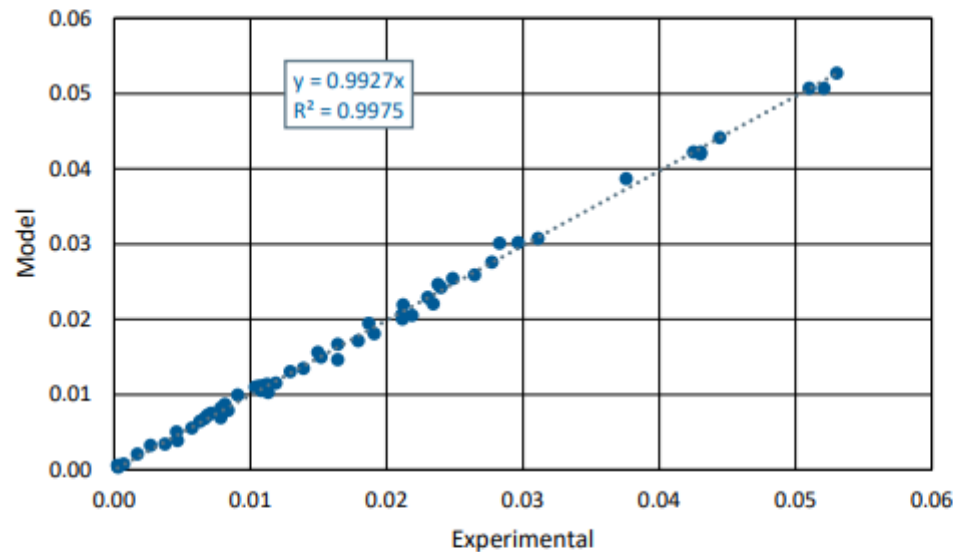
Designation	Characteristics	NaCl rejection
RO90	Thinfilm composite	≥ 90%**

** measured on 2000 ppm NaCl, 9 bar, 25°C

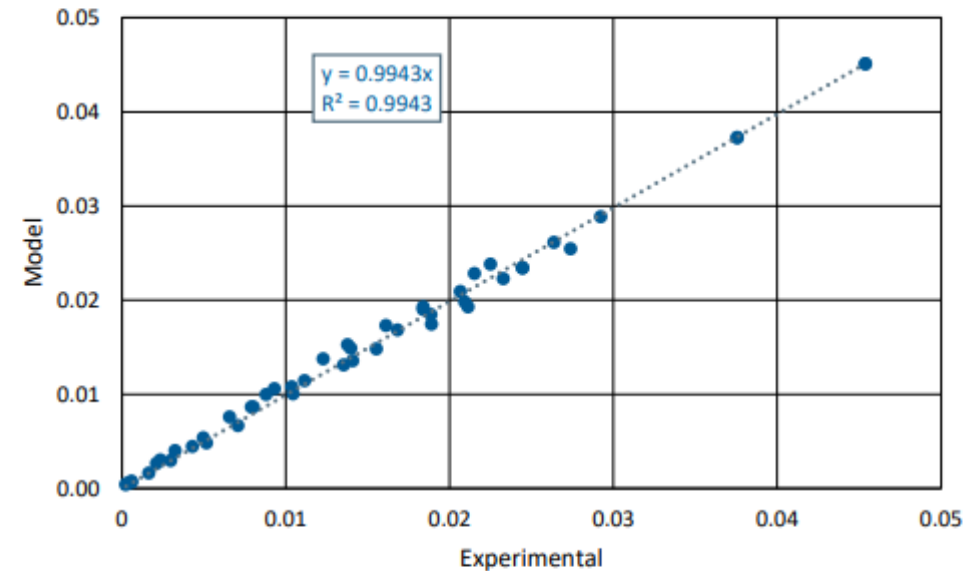
Part No.	Element type	Area	
		m2	(sqft)
527977	RO90-8038/30	38	(409)
525510	RO90-8038/48	28.1	(302)



Parity plot RO90-8038/48



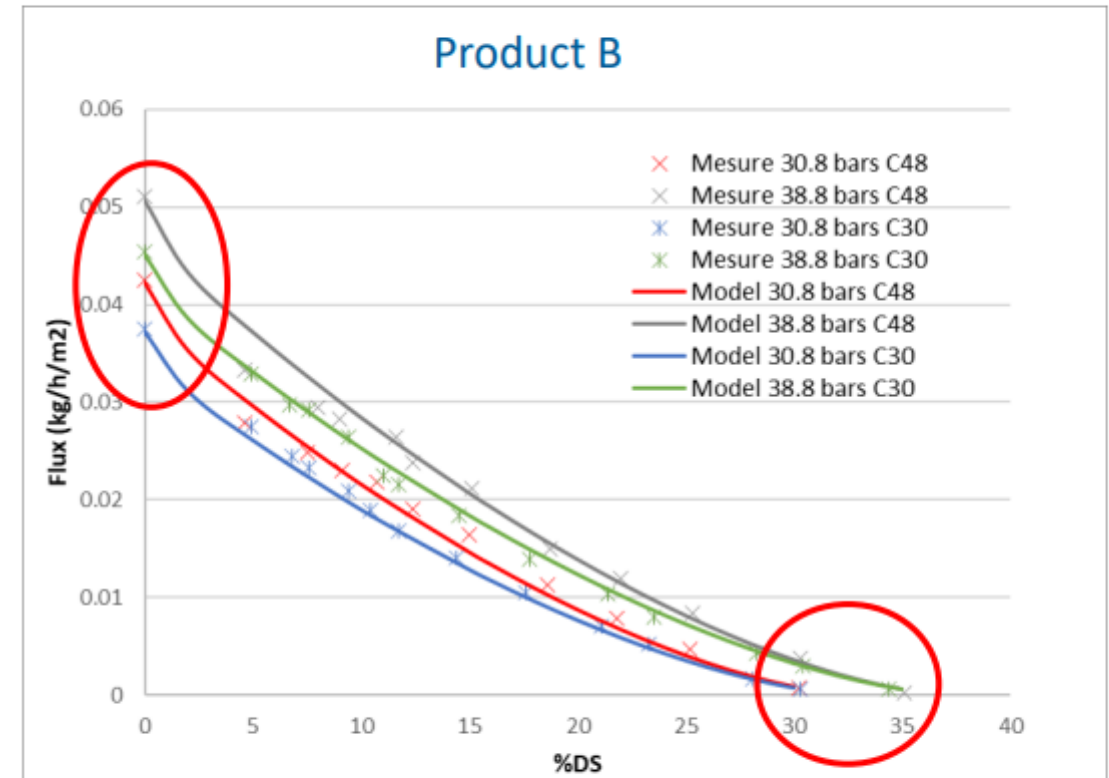
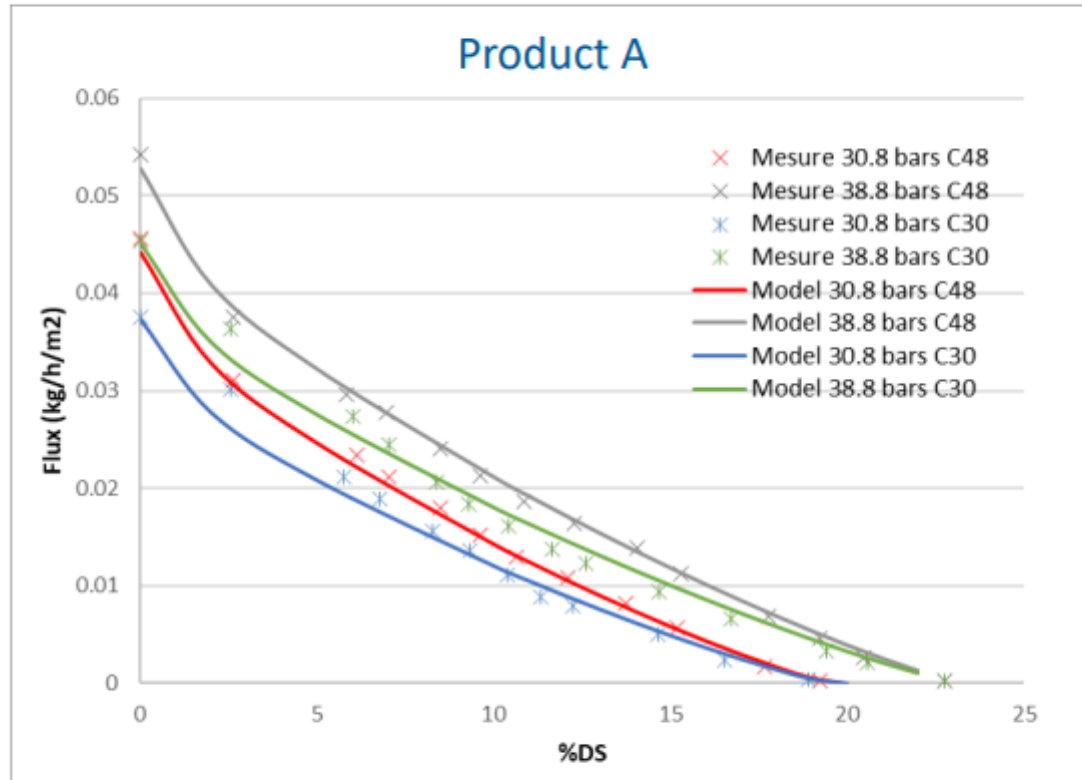
Parity plot RO90-8038/30



FLUX MODELING

$$J_p = L_p \cdot \frac{P_{TMP}}{\left(1 - \sigma \cdot \left(\frac{\Delta\pi_m}{P_{TMP}}\right)^p\right)^n}$$

Model	Model Osmolalityfalaval RO90-48		Model Osmolalityfalaval RO90-30	
Lp (kg/m²/Pa/s)	1.40E-08		1.20E-08	
Product	Product B	Product A	Product B	Product A
σ	0.835	0.924	0.835	0.924
p	0.521	0.566	0.524	0.571
n	-1.61	-1.20	-1.61	-1.20



REJECTION MODELING

$$f_{rej,i} = 1 - \frac{C_{p,i}}{C_{f,i}}$$

$$f_{rej,i} = \frac{1}{1 + a_i \cdot Jv^{b_i}}$$

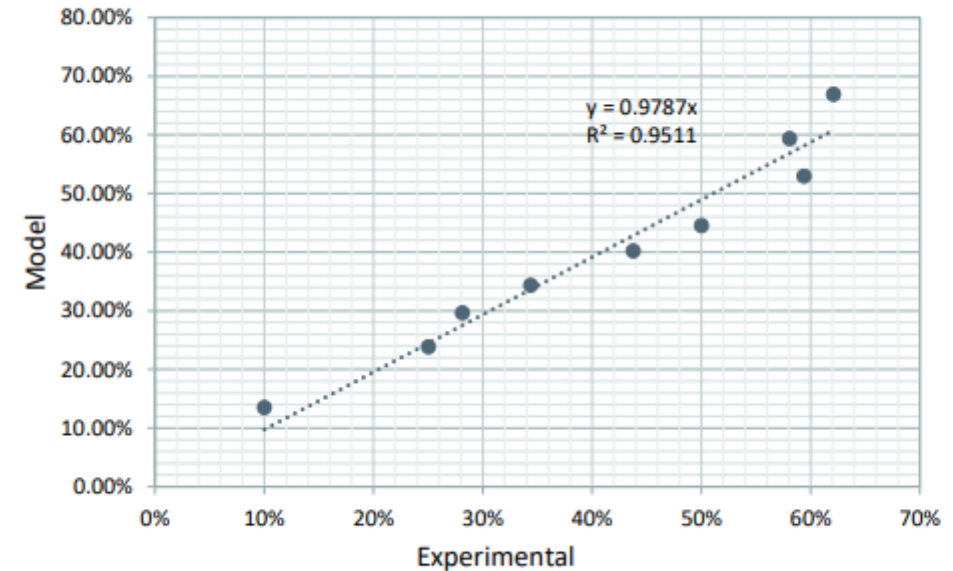
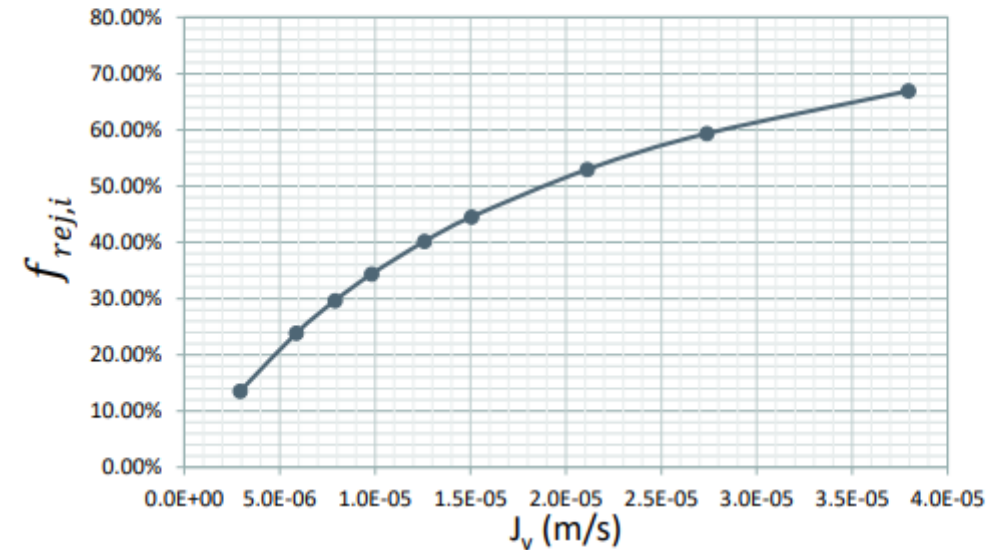
$C_{p,i}$ Mass concentration of component i in permeate stream $\left[\frac{kg}{m^3}\right]$

$C_{f,i}$ Mass concentration of component i in feed stream $\left[\frac{kg}{m^3}\right]$

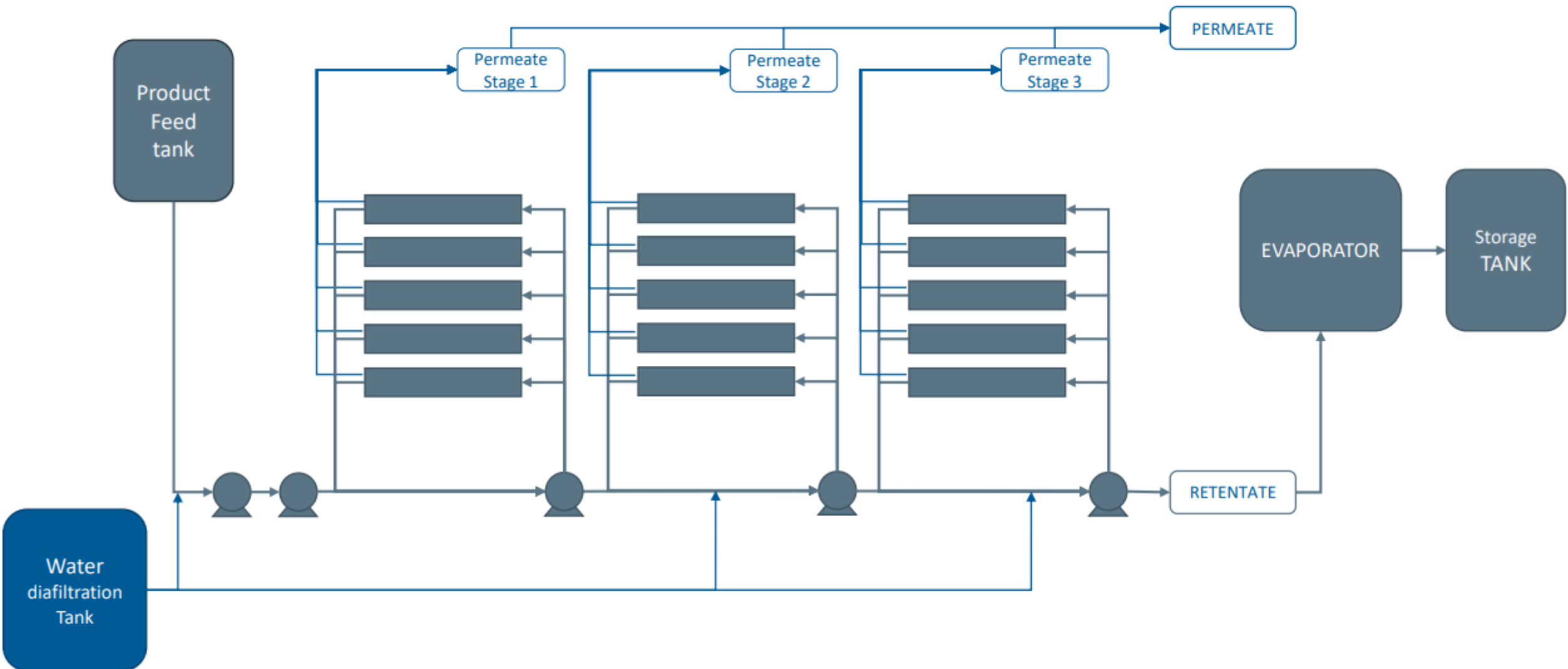
Jv Membrane permeate linear velocity $\left[\frac{m}{s}\right]$

a_i & b_i Rejection coefficient of component i at temperature T_i

Coefficient	Impurity	Product
a_i	1.8749E-05	0
b_i	-1	1



PROCESS OVERVIEW



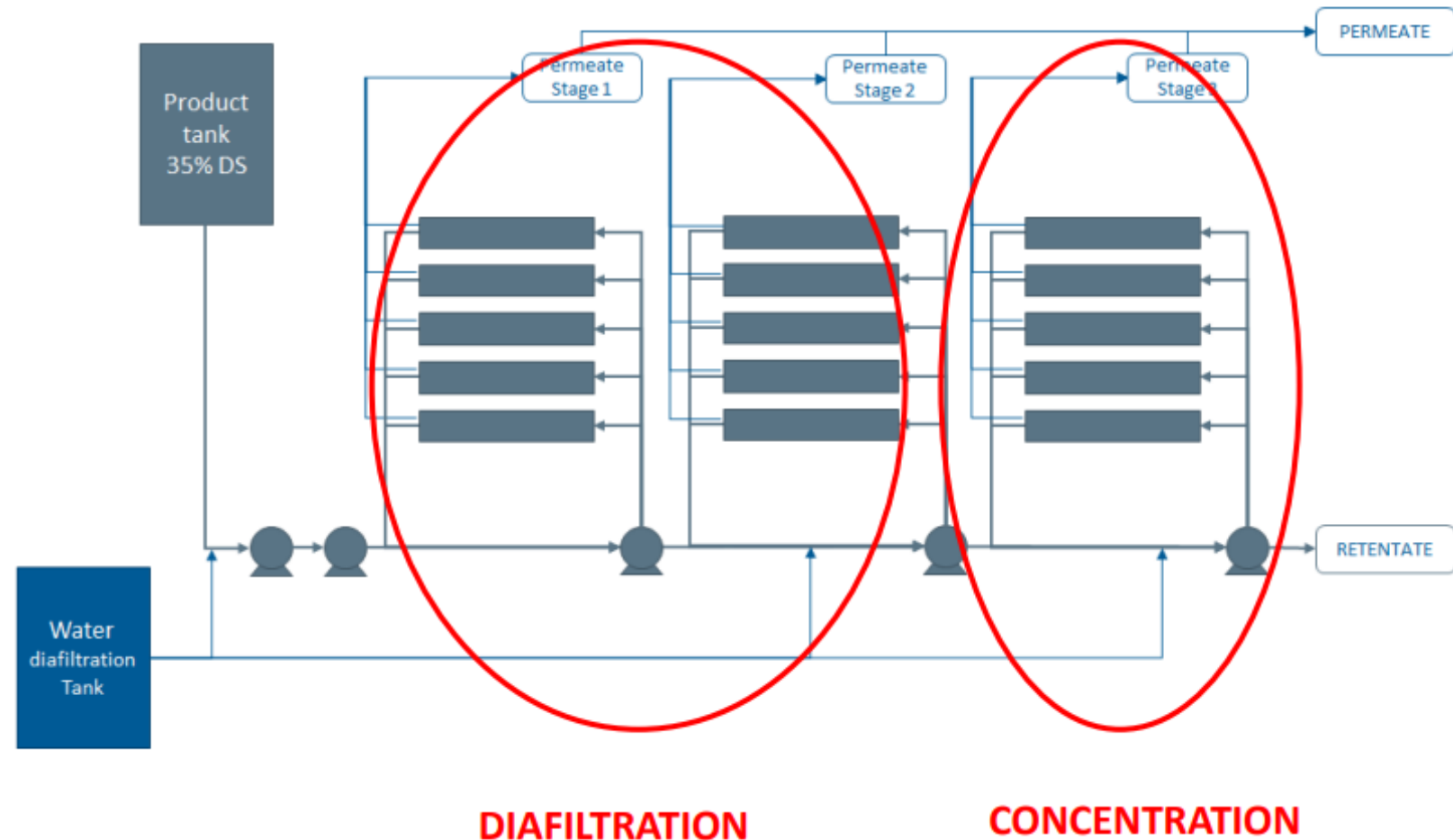
NANOFILTRATION CONTINUOUS

Continuous 3 stages of 5*4 membranes, total 60 membranes

Designation	Characteristics	NaCl rejection
RO90	Thinfil composite	≥ 90%**

** measured on 2000 ppm NaCl, 9 bar, 25°C

Part No.	Element type	Area	
		m2	(sqft)
527977	RO90-8038/30	38	(409)
525510	RO90-8038/48	28.1	(302)



PROCESS SIMULATION

gPROMS FormulatedProducts 2023.1.0

Projects: EG_industrial_continu_RO90_C48_30bars_VgPROM2023_1_0_VX

Models: Continuous_ALFALAVAL, Custom_membrane_rejection_model_gFP, Custom_membrane_static_resistance_model_gFP

Processes: Continuous_ALFALAVAL, Continuous_ALFALAVAL

Optimisation, Global System Analysis, Material Systems

Ultrafiltration_unit (Membrane_gFP)

Design mode: Off

Equipment and operation: Diafiltration split fraction per stage

Membrane resistance

Membrane selectivity

Diafiltration

Number of sta:	Number of diafiltration inlets
1	1
2	0
3	0

Retained species: Selected species

Selected species

Impurity	
Product	

Model for membrane selectivity: Rejection coefficient

Rejection coefficient model: Custom

The custom retention factor is defined in 'Custom_membrane_rejection_model_gFP'

Custom retention parameters

Custom parameter names

Selected sp:	a	b
Impurity	1.8749E-05	-1
Product	0	1

Static resistance

Static membrane resistance model: Custom empirical

The custom static resistance is defined in 'Custom_membrane_static_resistance_model_gFP'

Custom resistance parameters

Lp	1.40E-08
aw2	-2.134E-01
aw1	-1.496E-02
aw0	1
sigma_cst	0.835206
sigma_power	0.521744
Mw(kg/mol)	0
n	-1.61506

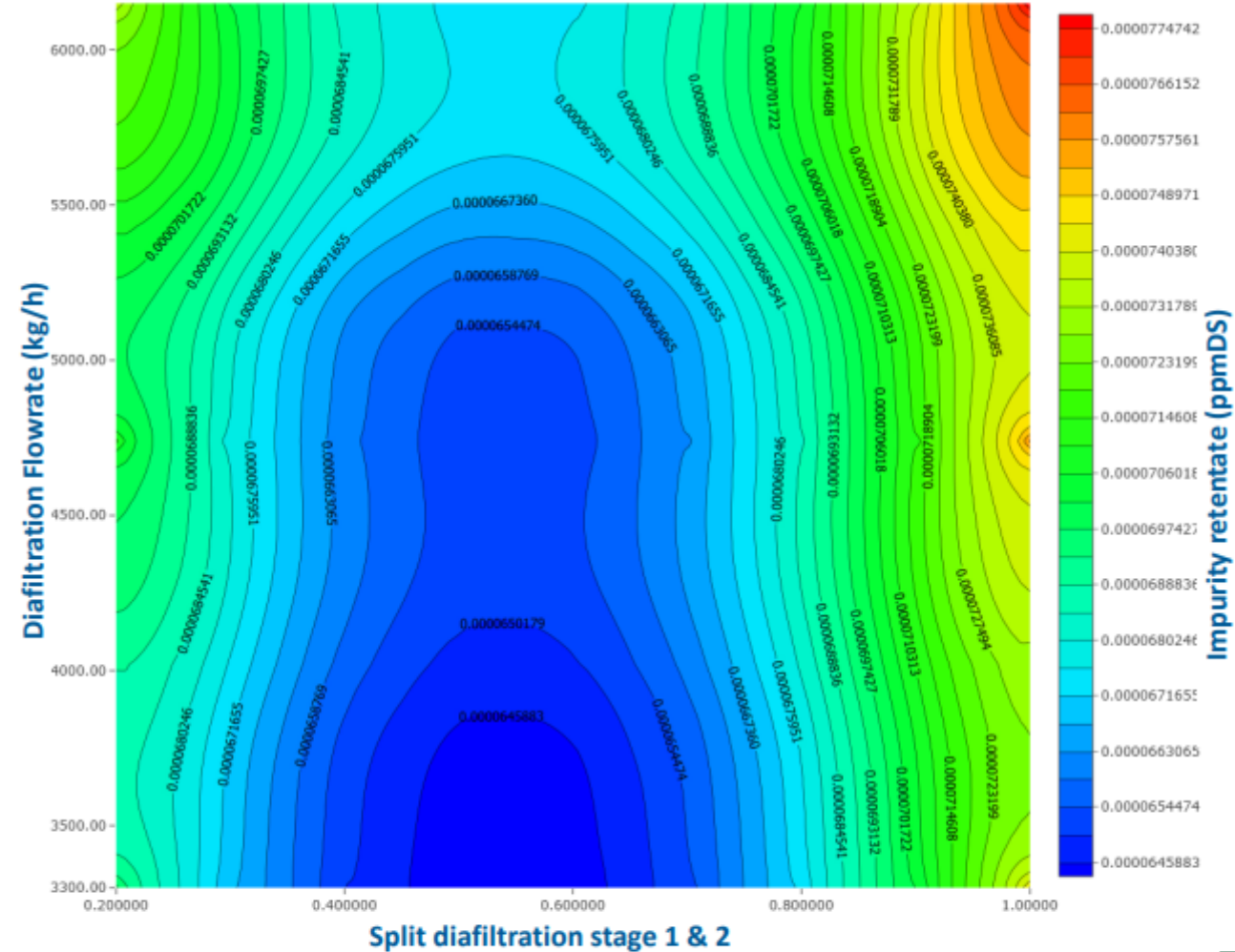
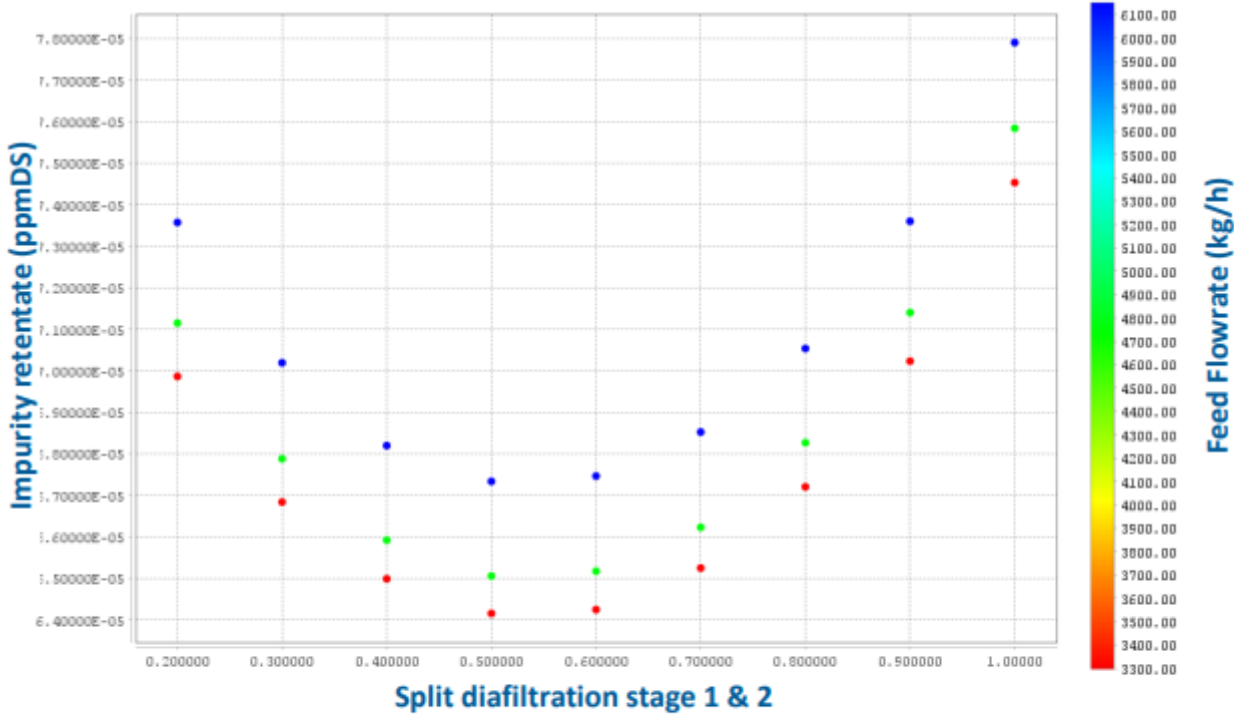
Dynamic resistance

Dynamic membrane resistance: Off

OK Cancel Reset all Help

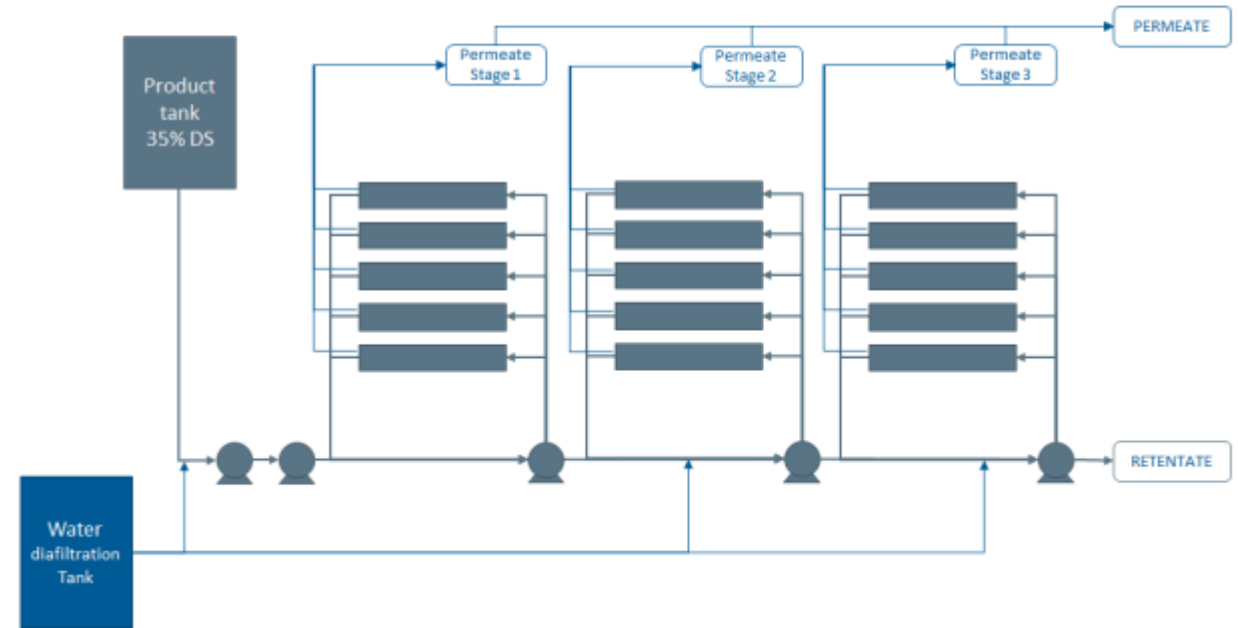
Rejection..._stage(3) -

PRODUCTIVITY OPTIMISATION - GSA



PRODUCTIVITY AND WATER CONSUMPTION OPTIMIZATION

Membrane surface (m2)	2280	TPM (bars)	29
Membrane	RO90-8038/30	m2	38.0
PRODUCT A			
Configuration split	1 on stage 1	0.4 stage 1	0.6 on stage 2
Productivity	2438 Tcom/year	8063 Tcom/year	
Productivity	0.23 TMS/h	0.75 TMS/h	
Water consumption	129.5 kgeau/kDS	24.7 kgeau/kDS	
Evaporation rate	0.9 T/h	3.0 T/h	
PRODUCT B			
Configuration split	1 on stage 1	0.5 stage 1	0.5 on stage 2
Productivity	18375 Tcom/year	17813 Tcom/year	
Productivity	1.72 TMS/h	1.66 TMS/h	
Water consumption	6.7 kgeau/kDS	4.4 kgeau/kDS	
Evaporation rate	3.0 T/h	3.0 T/h	



Split diafiltration water on stage 1 & 2 is more efficient

CONCLUSION

- New model approach for nanofiltration and reverse osmosis with osmotic pressure measurement method for complex product
- Easy implementation through custom model interface
- Parameters estimation tool for model validation
- GSA tool for scale-up and Quality by Design approach

In summary

- Food and beverage industries face a host of challenges including:
 - Cost pressures and supply chain uncertainty
 - Sustainability
 - Changing consumer demands
- Digitalization and digital twin creation is a growing trend
- Process modelling is a key part of digitalization which can help:
 - Scale up confidently
 - Improve process efficiency
 - Maintain product quality
 - Manage uncertainty

