



A simplified approach for activity monitoring in complex wastewater treatment processes

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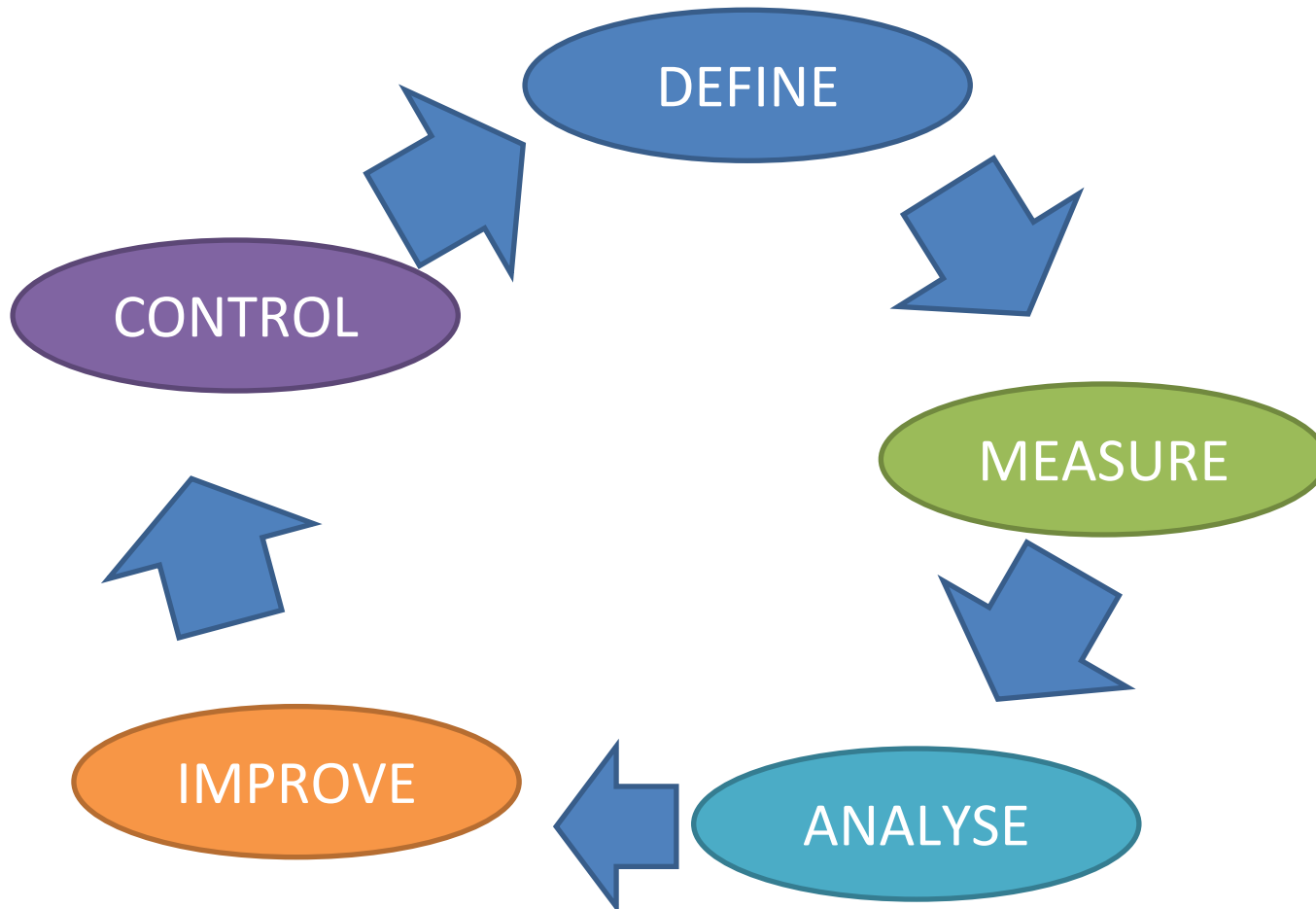
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Environmental Biotechnology Group - BIOGROUP

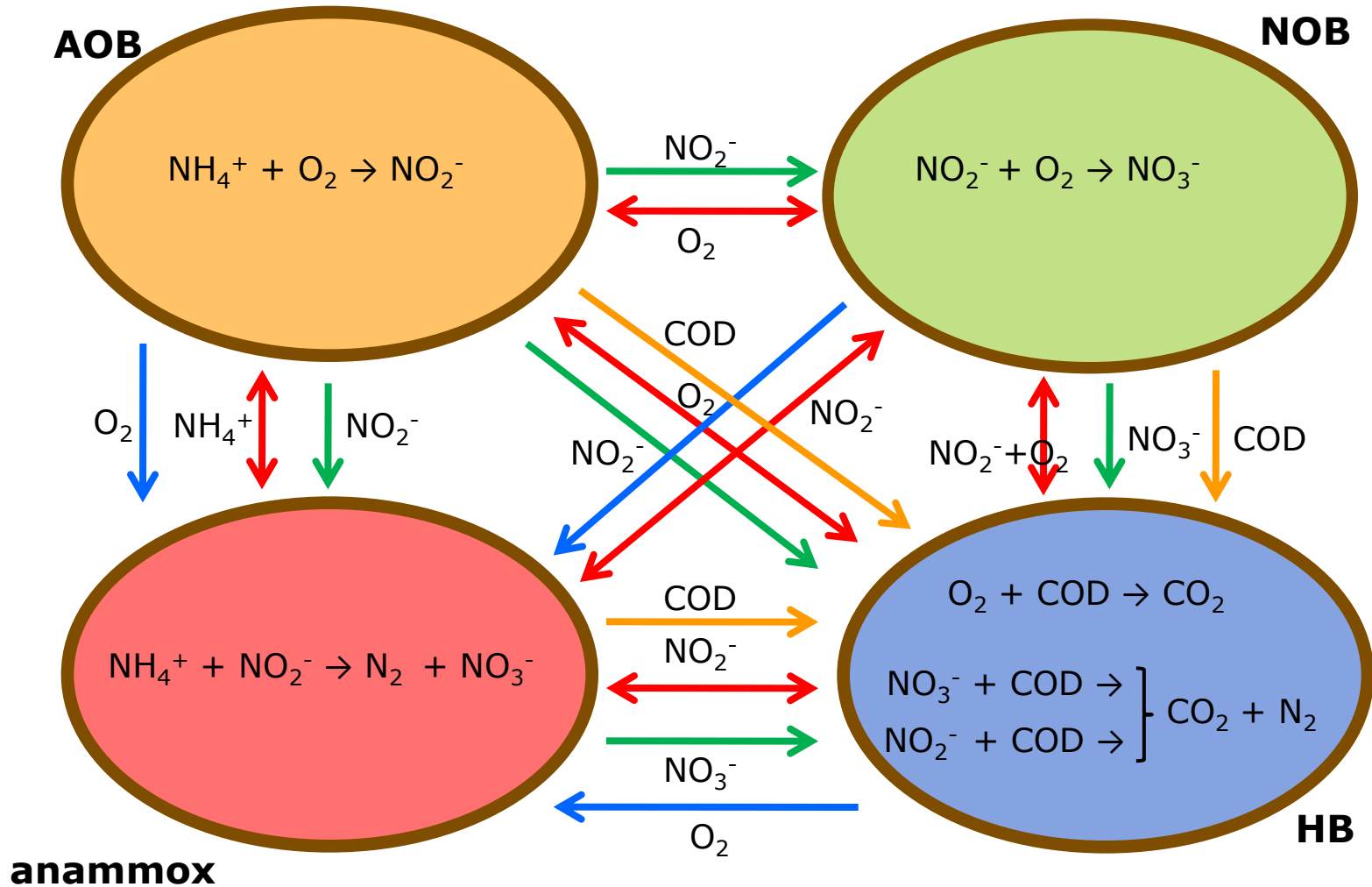
Universidade de Santiago de Compostela

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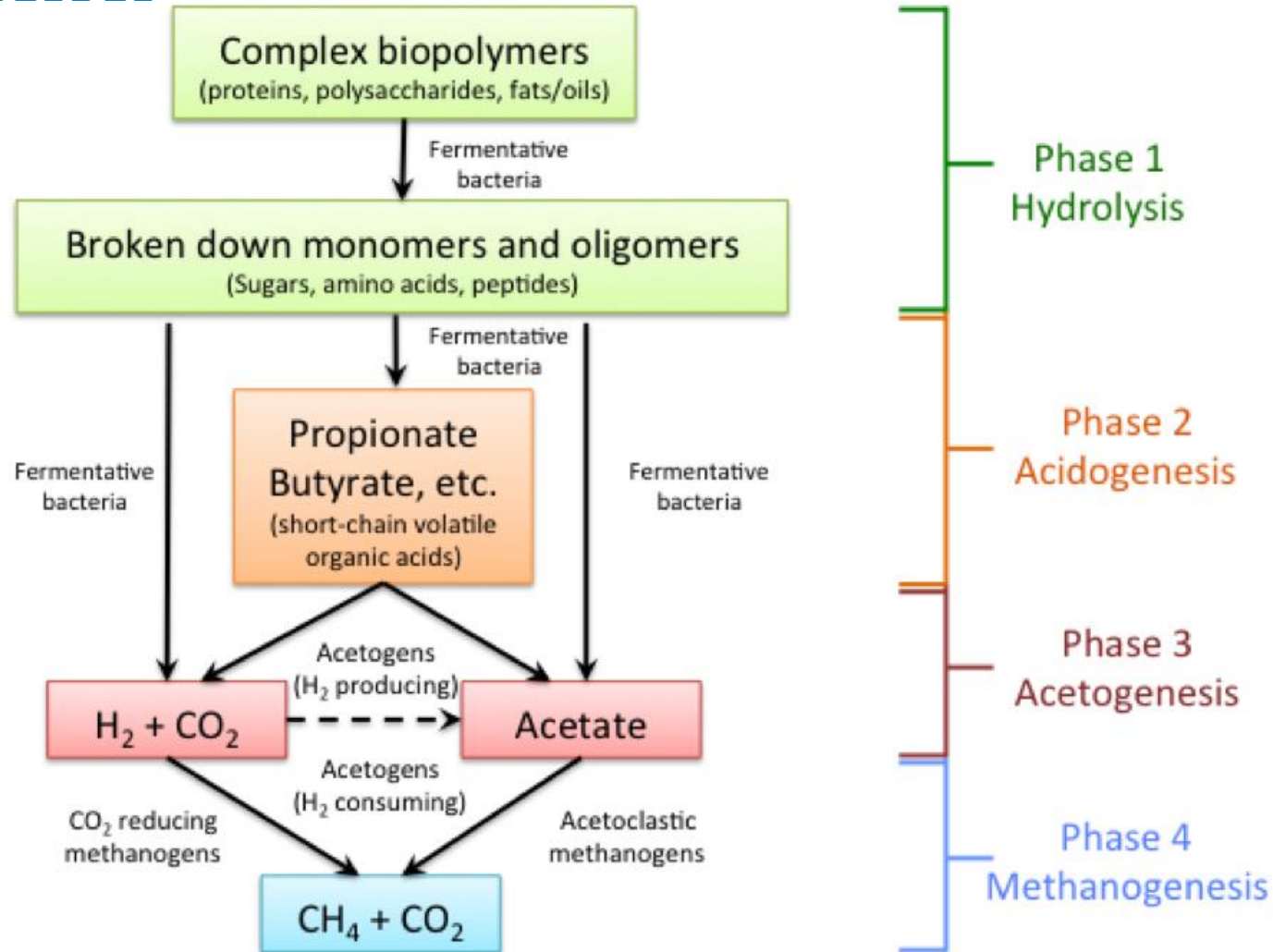
What you don't measure, you can't improve



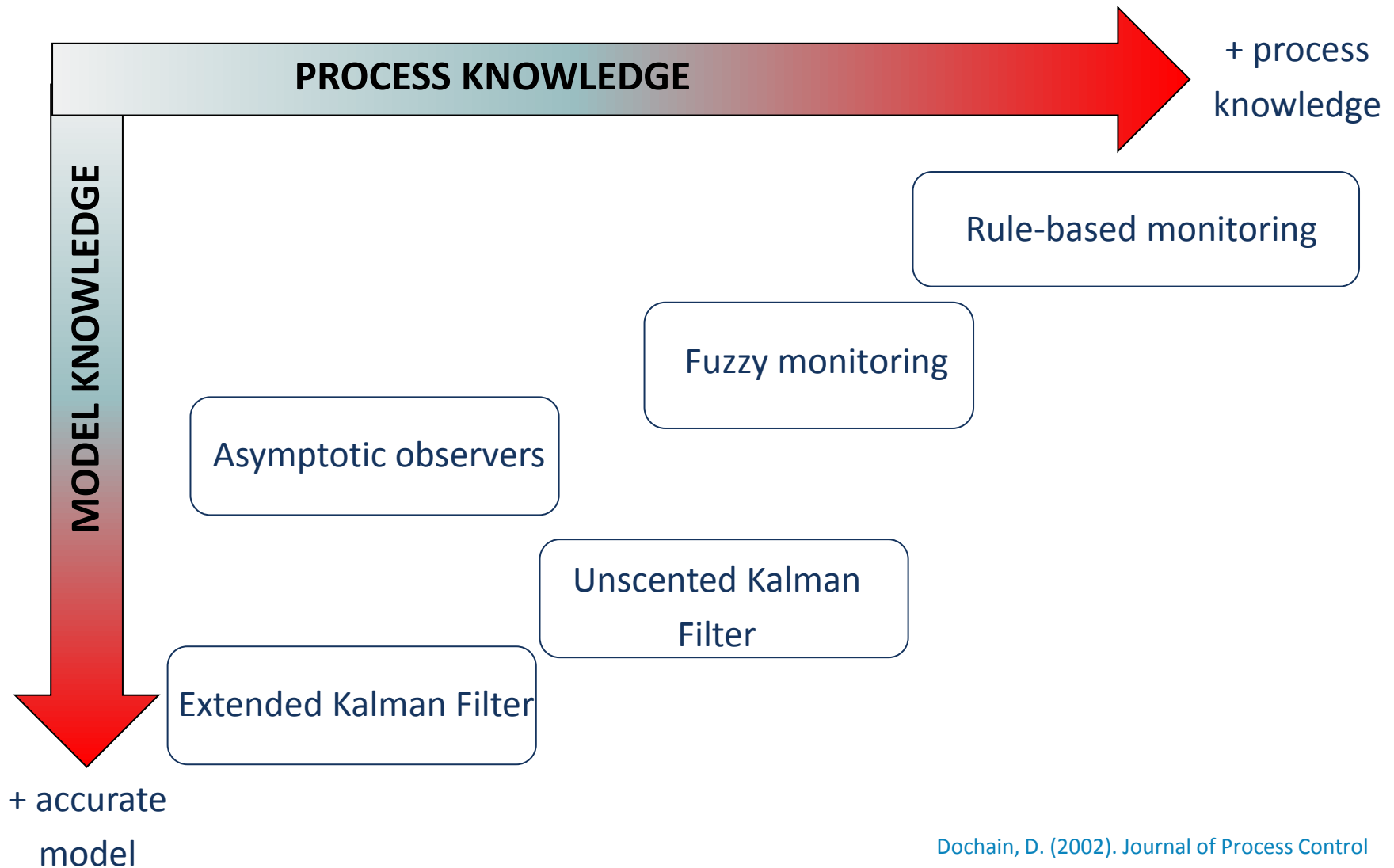
Activity rate monitoring is useful in some processes



Activity rate monitoring is useful in some processes



Most approaches rely on having a model or experience in running the process



The Continuous-Discrete Extended Kalman Filter is a very good tool if a model is available

Plant model

$$x_{k+1} = x_k + \int_{t_k}^{t_{k+1}} f(x(\tau), u_k, d_k) d\tau + w_k$$

$$y_k = G(x_k) + v_k, \quad v_k \in N_{iid}(0, R_v)$$



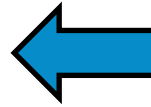
Measurement update

Error $e_k = y_k - C_k \hat{x}_{k|k-1}$

Gain $K_{f,k} = P_{k|k-1} C_k^T [C_k P_{k|k-1} C_k^T + R_v]^{-1}$

State estimate $\hat{x}_{k|k} = \hat{x}_{k|k-1} + K_{f,k} e_k$

Covariance $P_{k|k} = P_{k|k-1} - K_{f,k} [C_k P_{k|k-1} C_k^T + R_v]^{-1} K_{f,k}^T$



One step-ahead prediction

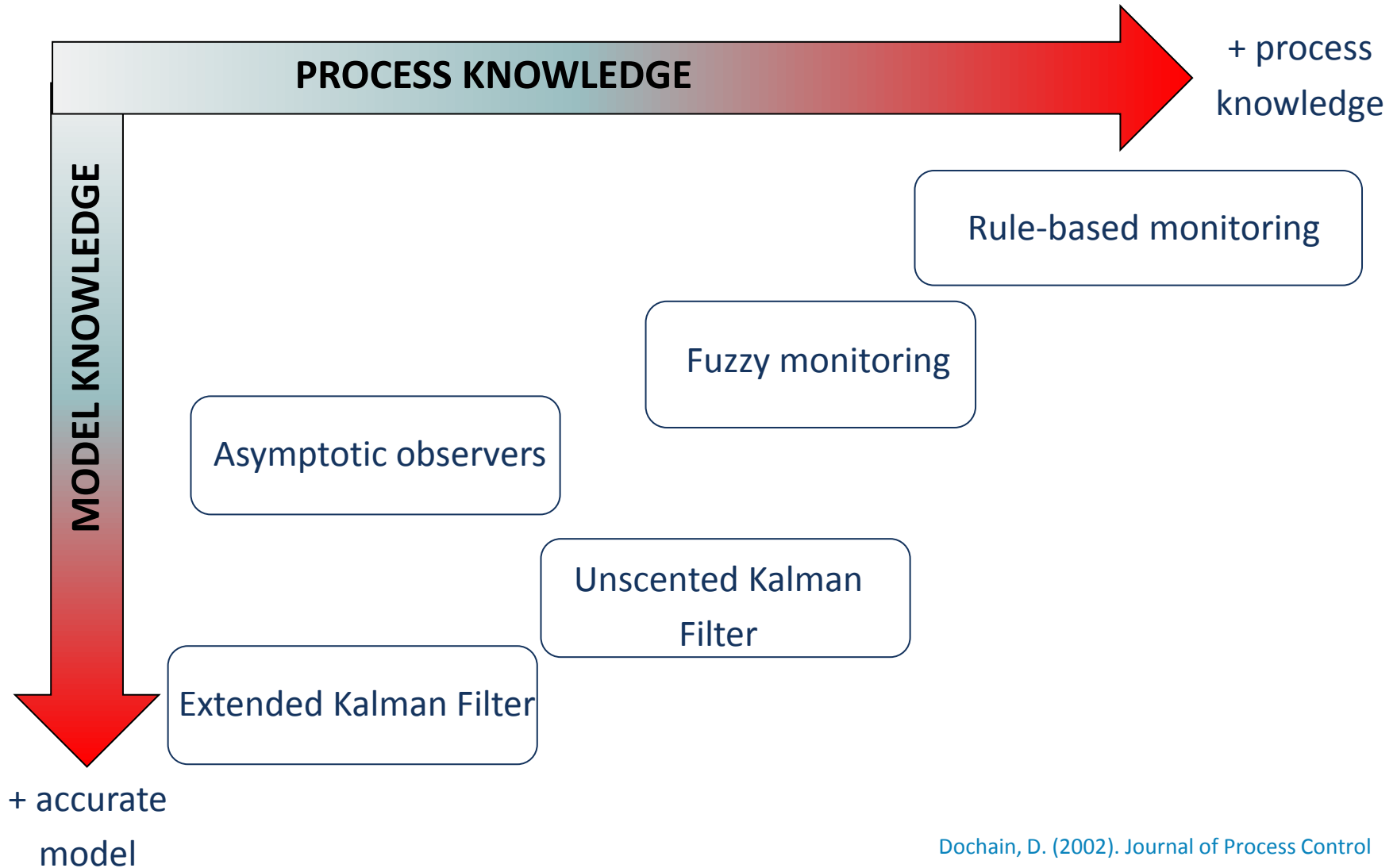
$$\hat{x}_{k+1|k} = \hat{x}_{k|k} + \int_{t_k}^{t_{k+1}} f(x(\tau), u_k, d_k) d\tau$$

$$\bar{A}_k = I + \int_{t_k}^{t_{k+1}} \frac{\partial f}{\partial x}(x(\tau), u_k, d_k) S_{\hat{x}_{k|k}} d\tau$$

$$P_{k+1|k} = \bar{A}_k P_{k|k} \bar{A}_k^T + R_w$$



What if we have little knowledge?



We propose stoichiometry-based monitoring as a very simple tool

$$r_S \approx \frac{S_t - S_{t-1}}{\Delta t} - F_{IN} S_{In} + F_{Out} S_t$$

Reaction = Accumulation - In + Out

This we know
(partially)

$$r_S = S \rho$$

This we want
to find out

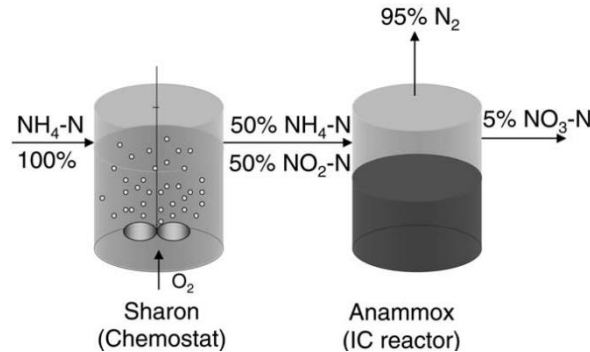
$$r_S = S \rho$$

More r_s than $\rho \rightarrow$ The system is **overdetermined** \rightarrow **Data reconciliation**

As many r_s as $\rho \rightarrow$ The system is **determined** \rightarrow **Unique solution**

Fewer r_s than $\rho \rightarrow$ The system is **underdetermined** \rightarrow **Infinite solutions**

A Sharon-Anammox process without HB gives place to a determined system of equations

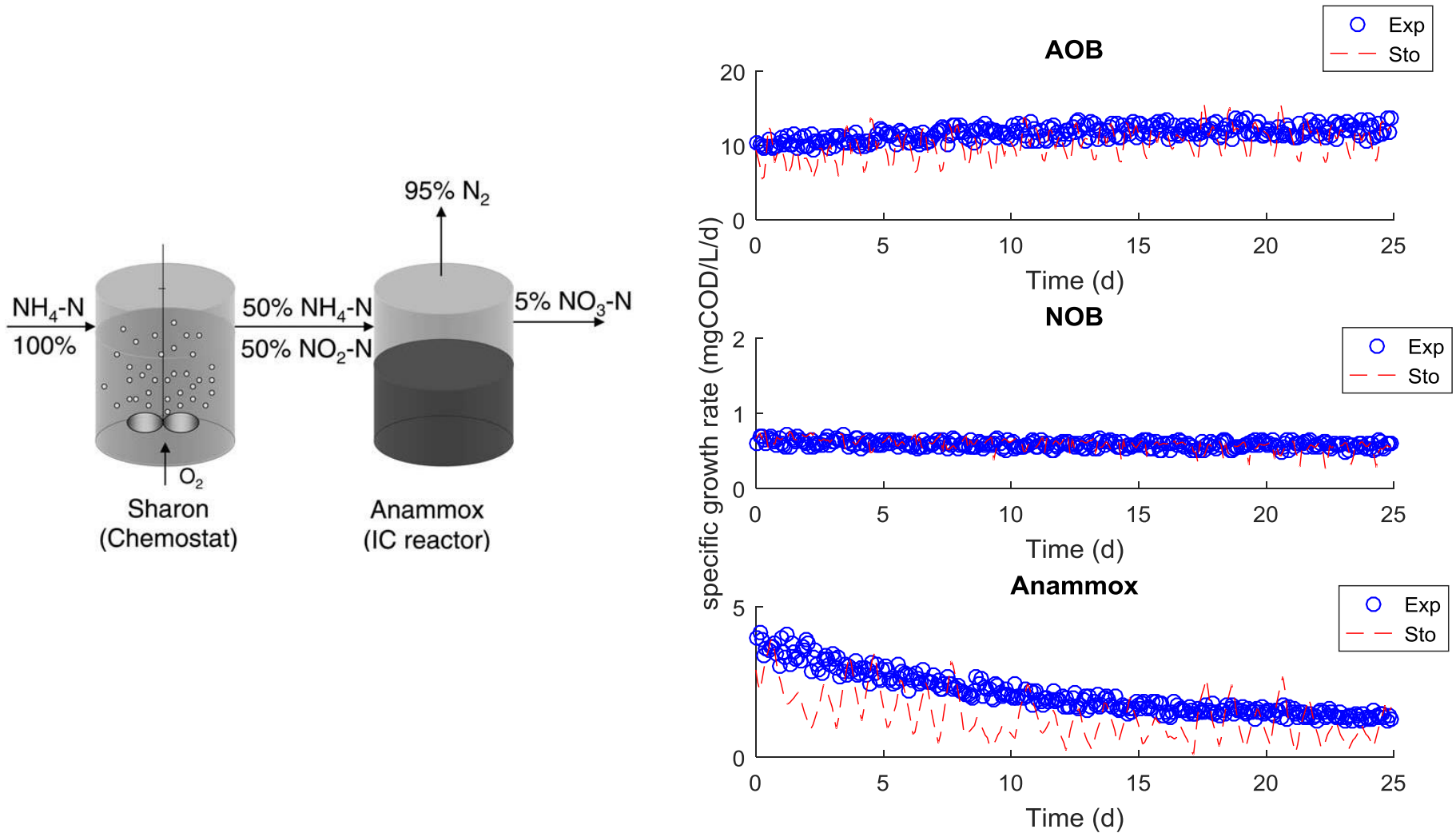


$$\begin{matrix} r_{NH} \\ r_{NO2} \\ r_{NO3} \end{matrix} = \begin{pmatrix} -1/Y_{AOB} & 0 & -1/Y_{Amx/NH} \\ 1/Y_{AOB} & -1/Y_{NOB} & -1/Y_{Amx/NO2} \\ 0 & 1/Y_{NOB} & 1.52 \end{pmatrix} \begin{pmatrix} \rho_{AOB} \\ \rho_{NOB} \\ \rho_{Amx} \end{pmatrix}$$

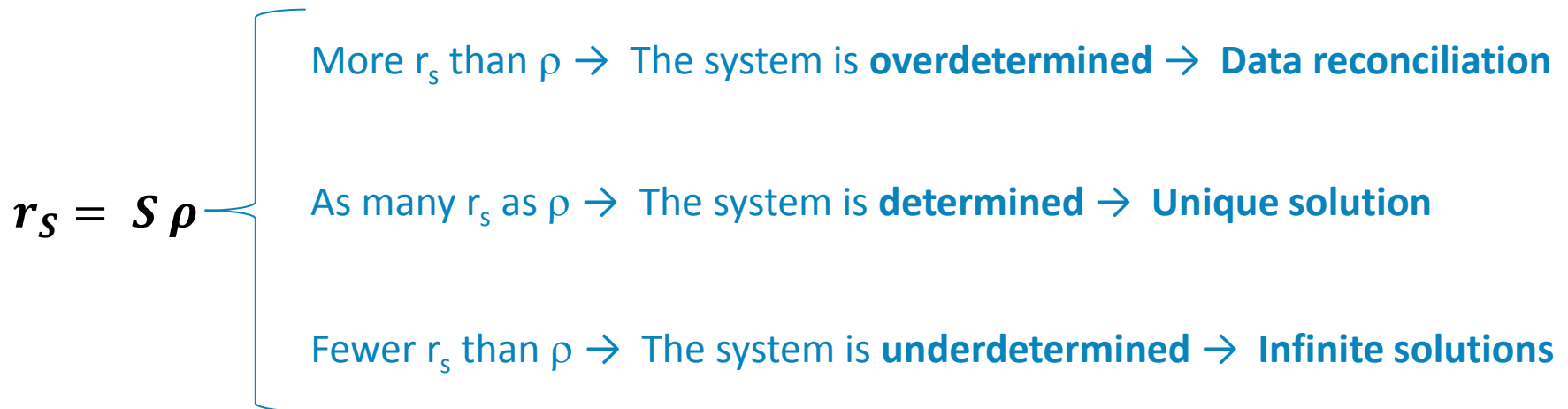
This we know
(partially)

This we want
to find out

Stoichiometry-based monitoring reproduces the reaction rates in an Sharon-Anammox process



In many practical cases, the system is underdetermined...



In many practical cases, the system is underdetermined...

$$r_s = S \rho$$

More r_s than ρ → The system is **overdetermined** → **Data reconciliation**

As many r_s as ρ → The system is **determined** → **Unique solution**

Fewer r_s than ρ → The system is **underdetermined** → **Infinite solutions**



Activity rates cannot take any value. They are non-negative

$$\rho \in \mathbb{R}_{\geq 0}$$

We turn an underdetermined system of equations into a linear least squares program

$$r_S = S \rho \quad \longrightarrow \quad \min_{\rho} \frac{1}{2} \|S \rho - r_S\|^2$$

And we enforce non-negativity by constraining the value of ρ

$$s. t. \quad \rho_i \geq 0 \text{ and } \rho_{inf} \leq \rho_i \leq \rho_{sup}$$

Stoichiometry-based monitoring only requires substrate/product stoichiometry

Step 1

Determine the stoichiometry matrix (\mathbf{S}), at least for substrate(s)/product(s) relationships



Step 2

Estimate the vector of observable rates (\mathbf{r}) based on macroscopic measurements by mass balances



Step 3

Solve the constrained linear least-squares problem

$$r_S \approx \frac{S_t - S_{t-1}}{\Delta t} - F_{IN} S_{In} + F_{Out} S_t$$

$$\min_{\rho} \frac{1}{2} \|\mathbf{N}\rho - \mathbf{r}\|^2$$

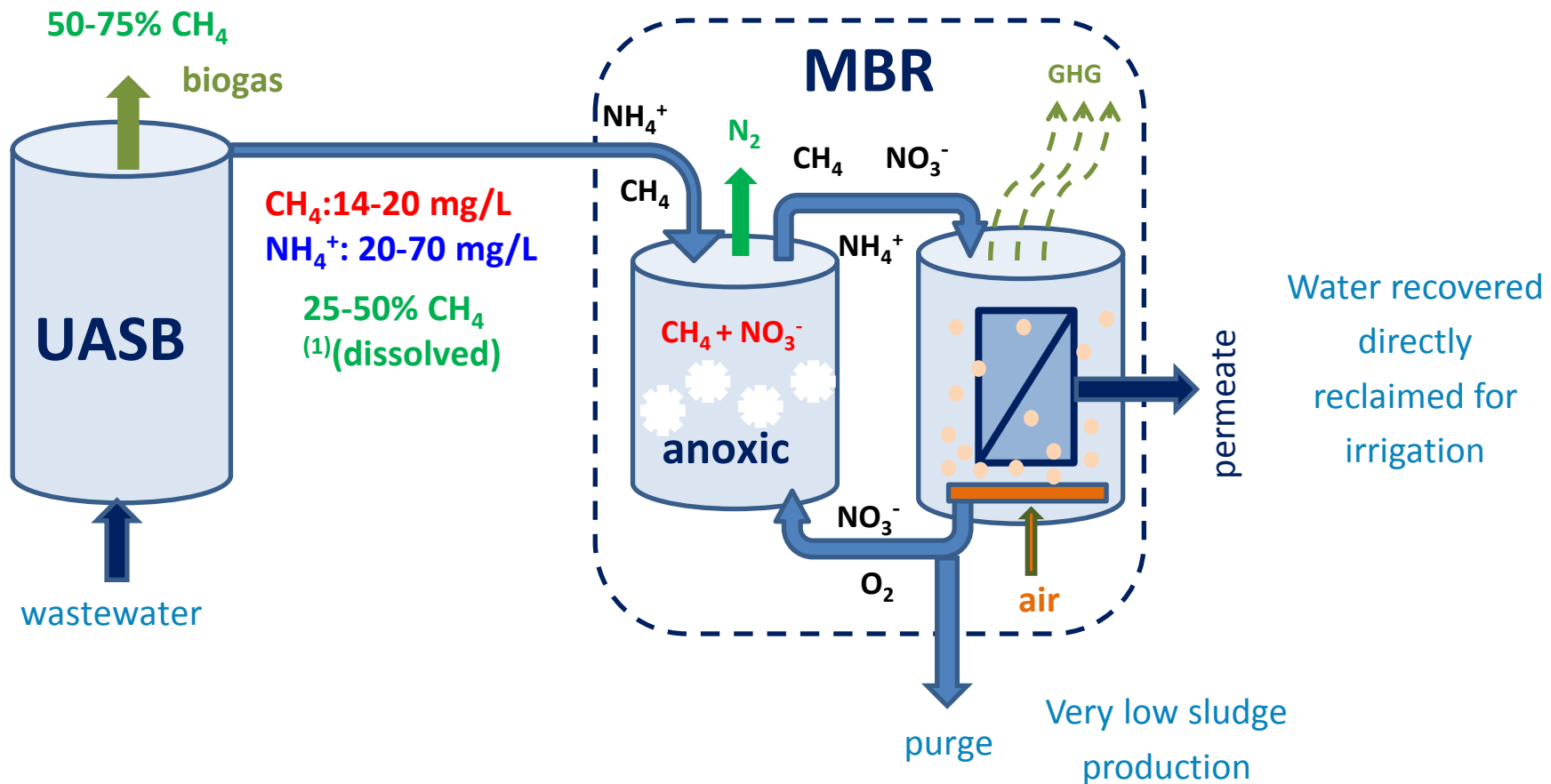
s. t. $\rho_i \geq 0$

and $\rho_{inf} \leq \rho_i \leq \rho_{sup}$

Case study. Novel SIAM Process

SIAM (*Sistema Integrado Anaerobio de Membrana*)

Membrane Integrated Anaerobic System. Buntner, D. et al(2011) Water Sci.Tech, 64(2), 397-402



(1) Noyola, A. Morgan, J.M., Lopez, J.E. (2006). Reviews in Environmental Science and Bio-Technology 5, 93-114.

SIAM was patented at USC and demo plants are being built in Spain

Santiago de Compostela

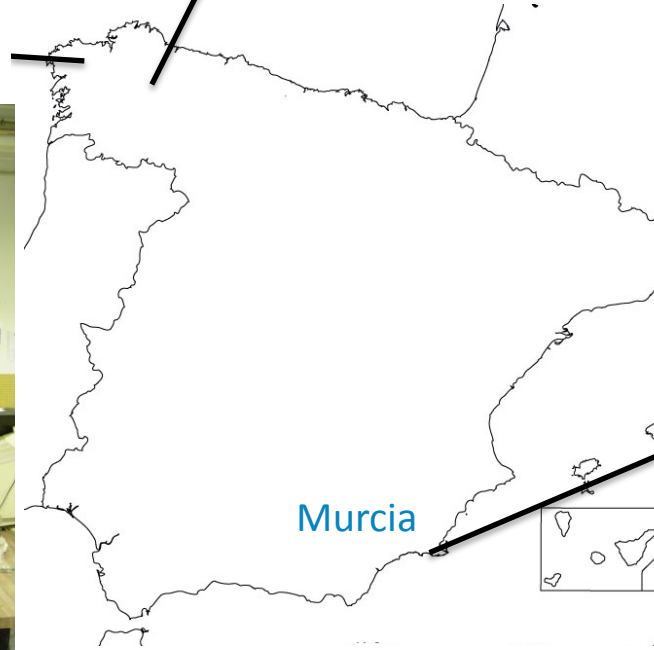
Anaerobic reactor (UASB)



Aerobic membrane bioreactor



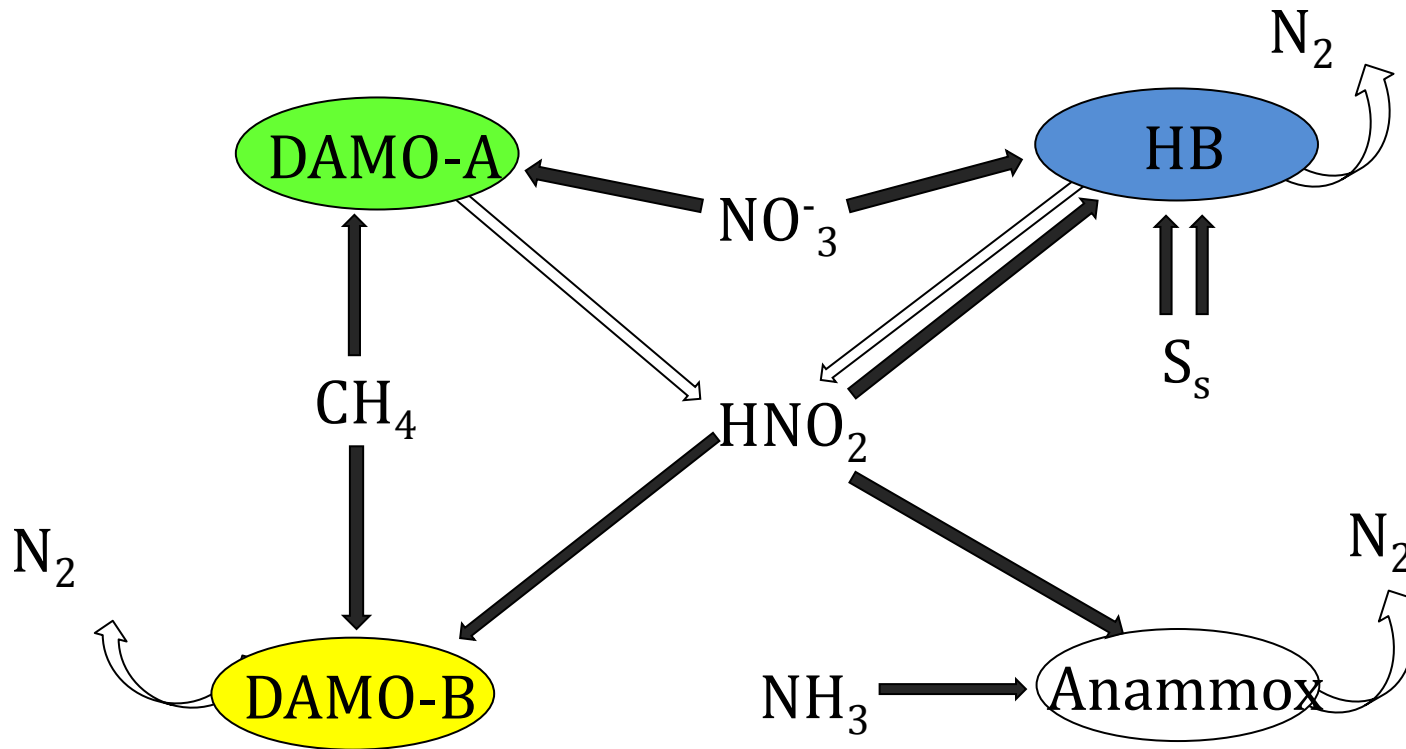
Outeiro de Rei (Lugo)
New plant in construction
in dairy company



Murcia

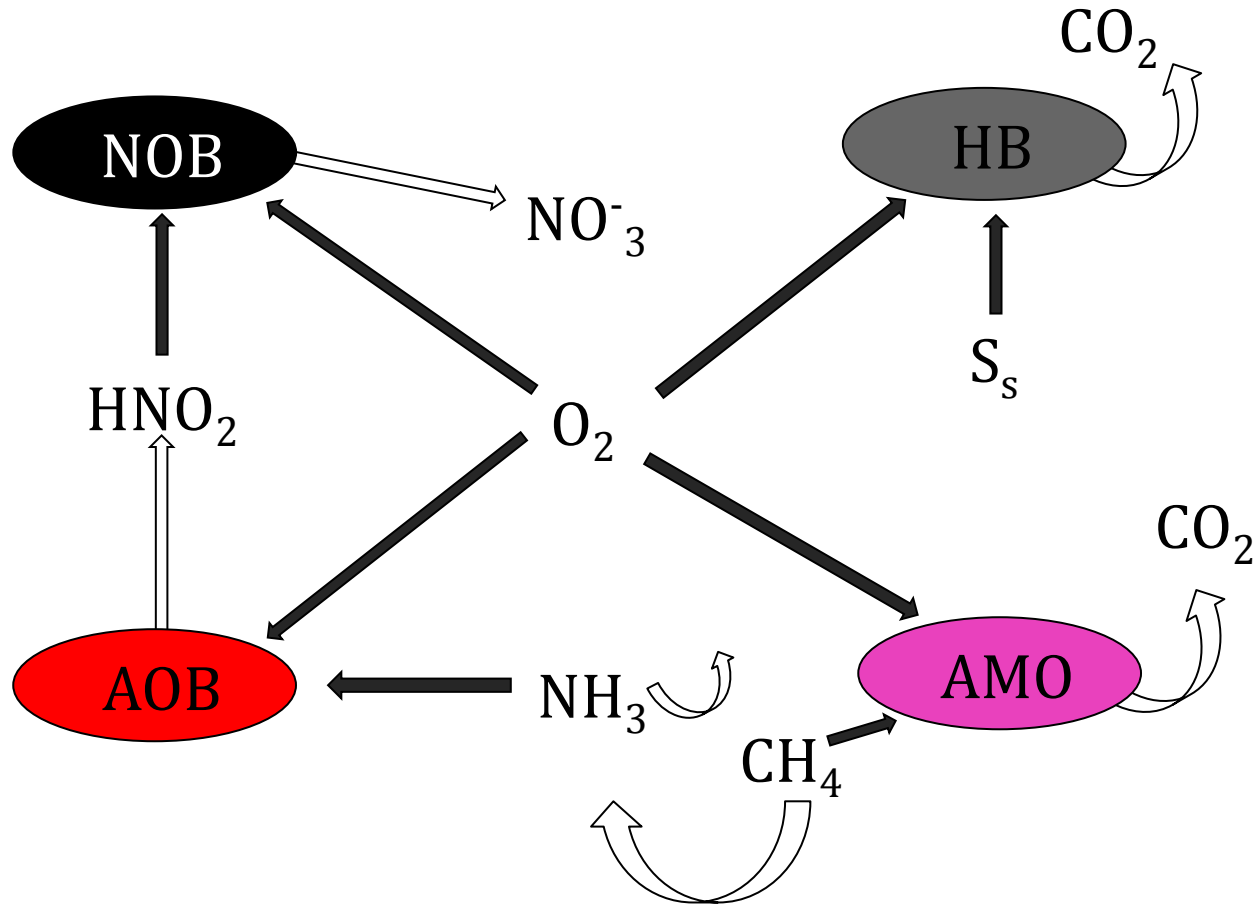


Many microbial groups play an important role in SIAM



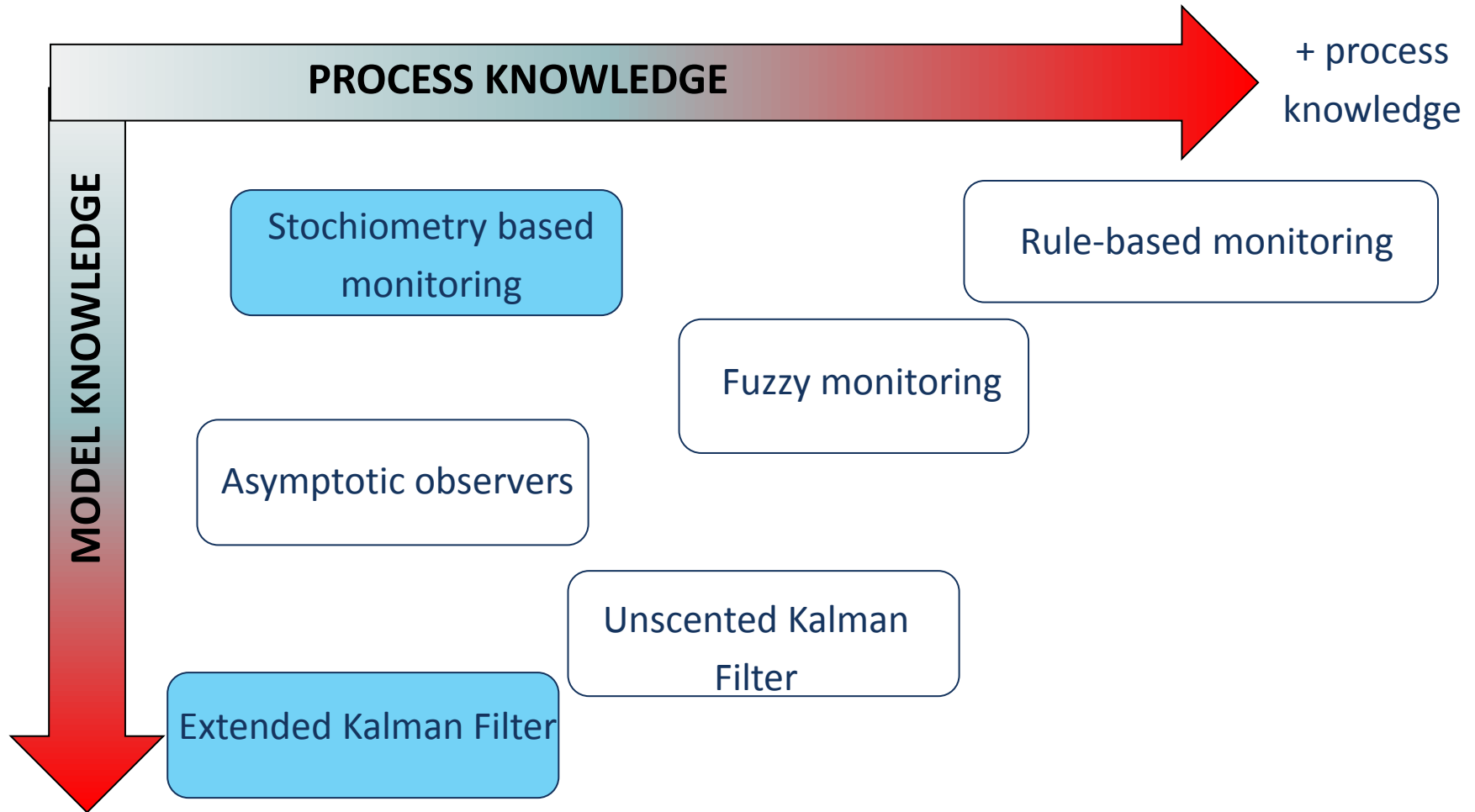
Main reactions in anoxic environment

Many microbial groups play an important role in SIAM



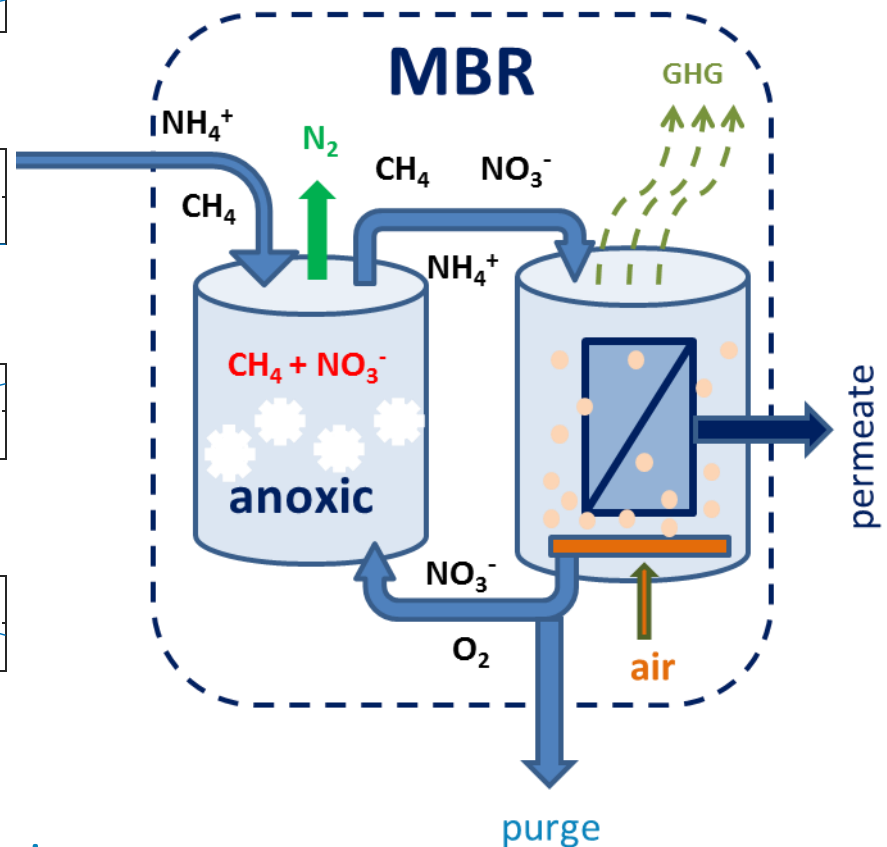
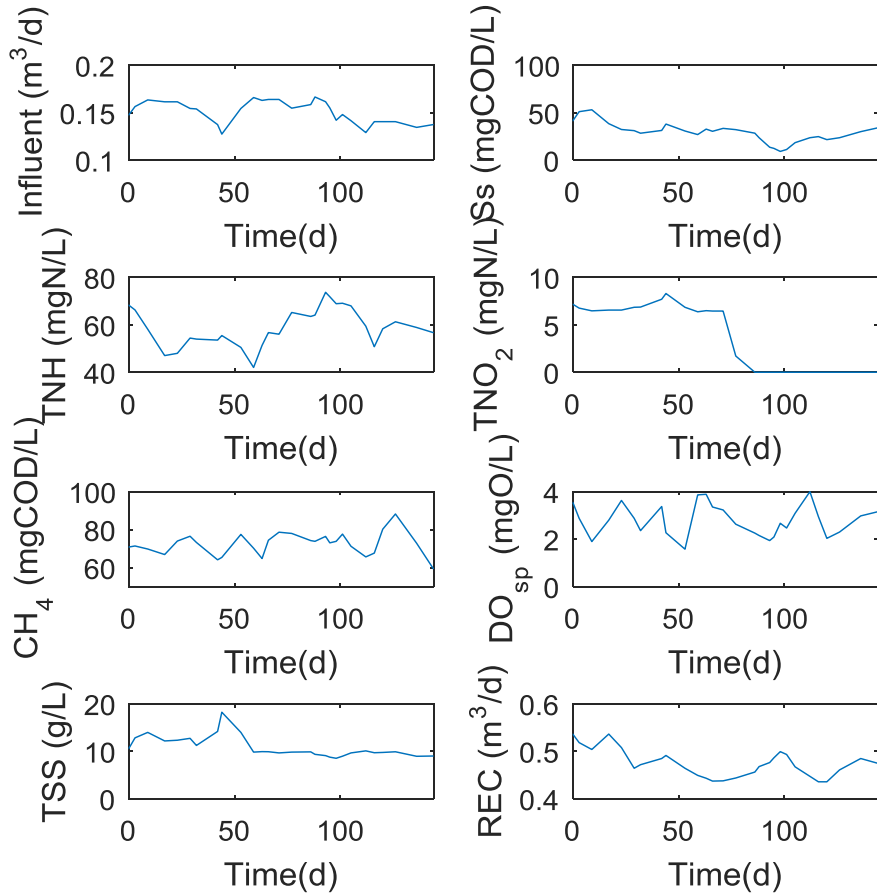
Main reactions in aerobic environment

Let's benchmark the new method with the continuous-discrete Kalman filter



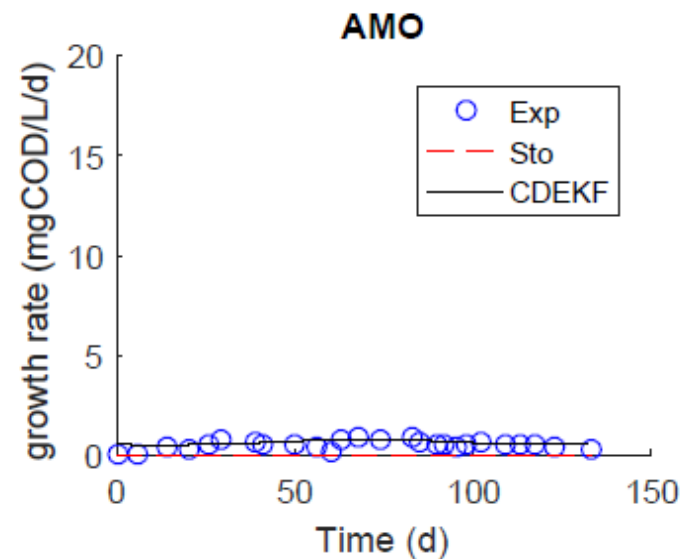
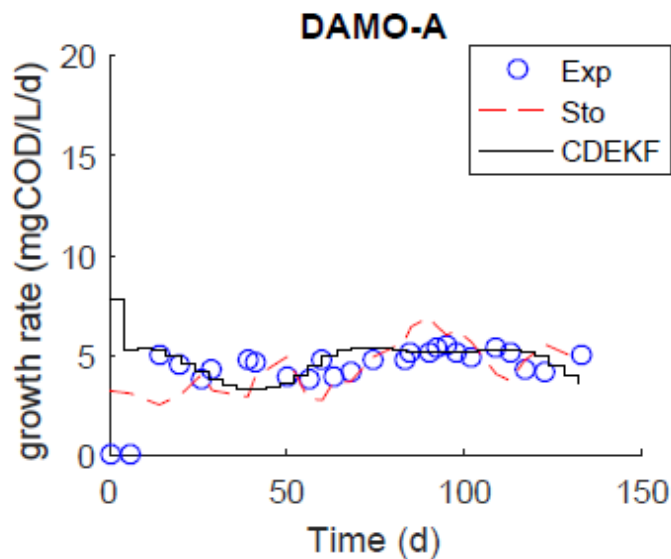
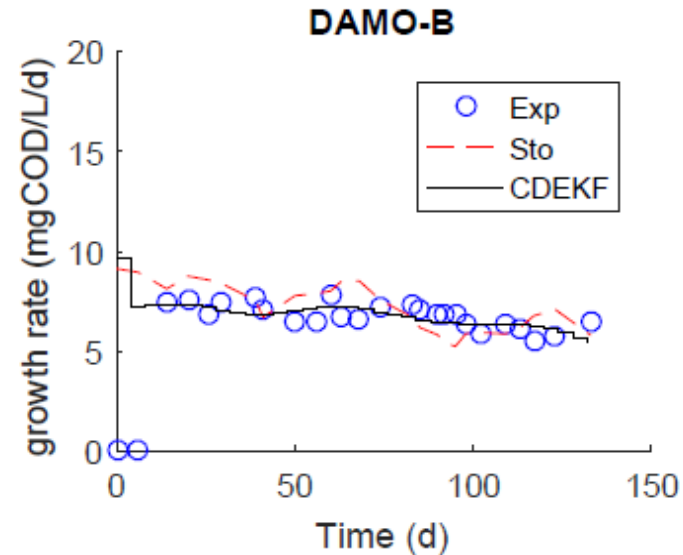
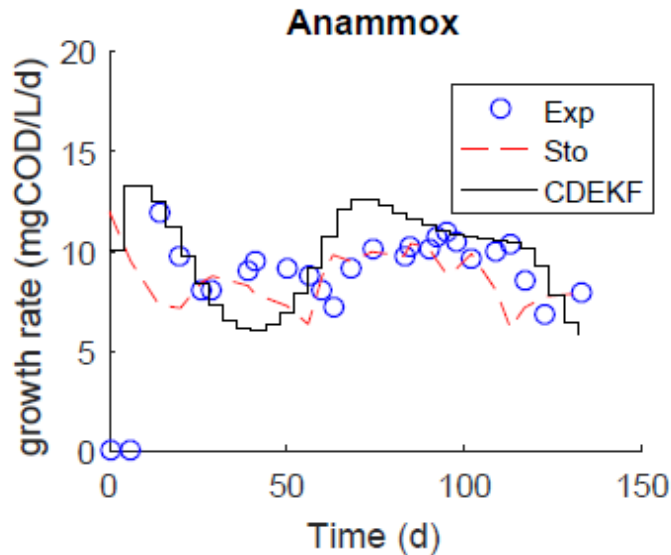
+ accurate process
knowledge

Using a realistic influent from our lab measurements

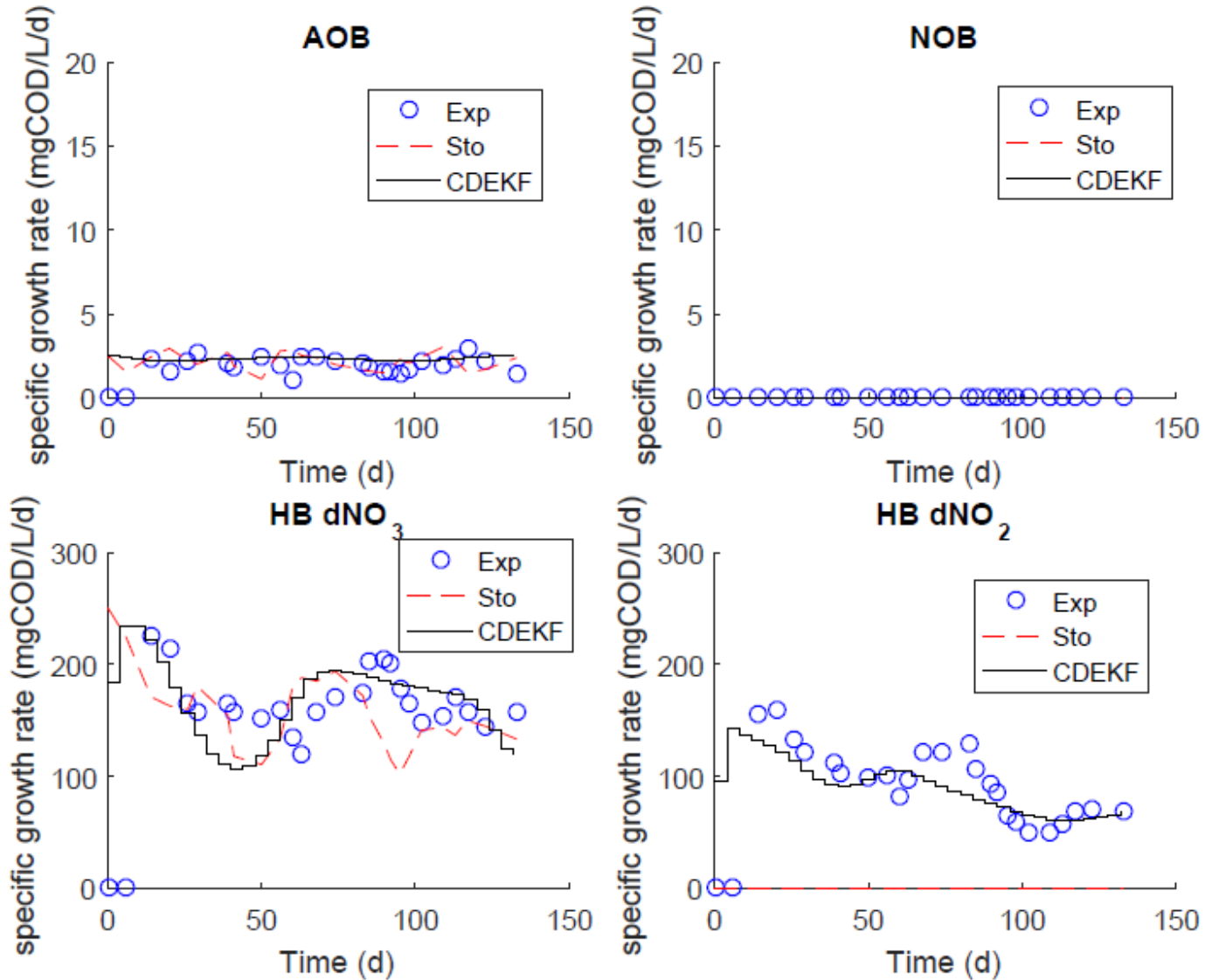


Measurements of ammonia, nitrite, nitrate, methane and oxygen in both chambers

The prediction of the rates is comparable for both methods ...



... except for heterotrophic denitrification



A promising method and a lot of work to develop it further

Can it be extrapolated to other processes?

Unicity?

What is the minimum number of measurements?

Appropriate sampling rate?



Simple



Intuitive



Limited amount of data



Easy to program



Surprisingly robust

Take home message

“There is no such a thing as simple methods;
there are methods adapted to a purpose” Gustaf Olsson (Academic’s Forum Sunday)

**The stoichiometry-based monitoring is well adapted for the
development of novel technologies**

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