

System Visualization Using Real-Time Data-Driven Models Derived from High-Resolution Sensor Profiling

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Background

- Wastewater/water treatment plant
- Develop physics-informed datadriven model to simulate the system heterogeneity
 - Conductivity
 - pH
 - Temperature
- Three sensors are deployed along the reactor



[1] Wang, C.; Wang, T.; Xu, Z.; Cui, C.; Wang, X.; Demitrack, Z.; Dai, Z.; Bagtzoglou, A .; Stuber, M. D.; Li, B., under review, 2020

[2] Xu, Z.; Dehkordy, F. M.; Li, Y.; Fan, Y.; Wang, T.; Huang, Y.; Zhou, W.; Dong, Q.; Lei, Y.; Stuber, M. D., High-fidelity profiling and modeling of heterogeneity in wastewater systems using milli-electrode array (MEA): Toward high-efficiency and energy-saving operation. Water research 2019, 165, 114971.

Conductivity Heterogeneity Profiling Model



Assumptions:

- Stirring power dominates the mass transfer
- Diffusion between adjacent zones are neglected

$$\frac{dC_i}{dt} = \frac{1}{4V} k_i (C_0 + C_4 - C_i), \qquad i = 1,2,3$$
$$\frac{dC_4}{dt} = \frac{1}{4V} k_4 (C_1 + C_2 + C_3 - 3C_4).$$



Global Optimization

$$\min_{\boldsymbol{p}\in\Pi\subset\mathbb{R}^{n_p}}\phi(\mathbf{x}(\mathbf{p},t_1),\ldots,\mathbf{x}(\mathbf{p},t_{N_t}))$$

s.t. $\dot{\mathbf{x}}(\mathbf{p},t) = \mathbf{f}(\mathbf{x}(\mathbf{p},t),\mathbf{p},t), \forall t \in I = [t_0,t_f]$
 $\mathbf{x}(\mathbf{p},t_0) = \mathbf{x}_0(\mathbf{p}).$

Objective function: ϕ (Optimization variable: \mathbf{p} =Generic state variable: \mathbf{x} =

$$\phi(\mathbf{x}(\mathbf{p}, t_1), \dots, \mathbf{x}(\mathbf{p}, t_{N_t}))$$
$$\mathbf{p} = (k_1, k_2, k_3, k_4, C_v)$$
$$\mathbf{x} = (C_1, C_2, C_3, C_4)$$

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pH Heterogeneity Profiling Model



[1] Wang, C.; Wang, T.; Xu, Z.; Cui, C.; Wang, X.; Demitrack, Z.; Dai, Z.; Bagtzoglou, A.; Stuber, M. D.; Li, B., under review, 2020

Temperature Heterogeneity Profiling Model

$$\frac{dT}{dt} = -\frac{UA_c}{V\rho C_p} (T - T_c)$$

$$T = T_c + (T_0 - T_c) \exp\left[-\frac{Ut}{L\rho C_p}\right]$$

$$T_{i \in \mathcal{T} \subset \mathbb{R} p \in \Pi \subset \mathbb{R}^{n_p}} \sum_{i=1}^{N} \sum_{j=1}^{3} (T_i - T_{i,j}^{data})$$
s.t.
$$T_i = T_c + (T_0 - T_c) \exp\left[-\frac{Ut_i}{h\rho C_p}\right], i = 1, ..., N$$

$$\frac{25}{24}$$

$$\frac{2}{0}$$

$$\frac$$

[1] Wang, C.; Wang, T.; Xu, Z.; Cui, C.; Wang, X.; Demitrack, Z.; Dai, Z.; Bagtzoglou, A.; Stuber, M. D.; Li, B., *under review*, 2020
[4] M. E. Wilhelm and M. D. Stuber. EAGO.jl: easy advanced global optimization in Julia. Optimization Methods and Software, pages 1–26, aug 2020

Optimal Control of Continuous Flow Nitrification Reactor

- Continuous flow nitrification reactor
- Inlet and outlet streams
 - High zone
 - Middle zone
 - Low zone
- Air diffuser
- Conductivity sensors
- Impeller
- Controller
 - PI Control
 - MPC
 - EMPC



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Modified Conductivity Model



Kinetics and Aeration

• Nitrification Process

 $2NH_4^+ + 3O_2 \rightarrow 2NO_2^- + 4H^+ + 2H_2O$ $2NO_2^- + O_2 \rightarrow 2NO_3^-$

Ammonium consumption rate:

$$r_{\text{NH}_{4}^{+}} = -r_{\text{AOB}} X_{\text{AOB}},$$

$$r_{\text{AOB}} = r_{\text{AOB,max}} \frac{c_{\text{NH}_{4}^{+}}}{K_{\text{SAOB}} + c_{\text{NH}_{4}^{+}}} + \frac{c_{\text{NH}_{4}^{+}}^{2}}{K_{\text{IAOB}}} \frac{c_{O}}{K_{\text{OAOB}} + c_{O}}$$

• Oxygen balance





[5] Sánchez, O., Aspé, E., Martí, M.C. and Roeckel, M. (2004) The Effect of Sodium Chloride on the Two–Step Kinetics of the Nitrifying Process. Water environment research 76(1), 73-80.
 [6] Judd, S. (2010) The MBR book: principles and applications of membrane bioreactors for water and wastewater treatment, Elsevier.

Proportional-Integral Control

$$u(t) = K_p e(t) - K_i \int_0^t e(t') dt'$$

• Control variable

u(t) = Q

• Error

 $e(t) = SP - C_1(t)$

• Tuning coefficients

MEA MEA MEA MEA MEA MEA MEA

 K_p, K_i



Model Predictive Control

• Primary advantage: multiple-input systems

$$\min_{u_{k},\dots,u_{k+M-1}} \sum_{i=1}^{3} \sum_{j=1}^{P} (SP - \hat{C}_{i,k+j})^2$$

- u : control variable
- SP : setpoint
- \hat{C}_i : model predicted output (DMC)
- P : prediction horizon
- M : control horizon





Economic Model Predictive Control



Economic Model Predictive Control



Results



Discharge & Energy



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Any questions?

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