

Int. J. of Computers, Communications & Control, ISSN 1841-9836, E-ISSN 1841-9844
Vol. VI (2011), No. 2 (June), pp. 367-374

Digital Control of a Waste Water Treatment Plant

R. Vilanova, J.D. Rojas, V.M. Alfaro

Ramón Vilanova, José David Rojas

Department de Telecomunicació i Enginyeria de Sistemes
Universitat Autònoma de Barcelona
08193, Bellaterra, Spain,
E-mail: ramon.vilanova@uab.cat, josedavid.rojas@uab.cat

Víctor M. Alfaro

Escuela de Ingeniería Eléctrica
Universidad de Costa Rica
San José, 11501-2060 Costa Rica.
E-mail: victor.alfaro@ucr.ac.cr

Abstract: The Activated Sludge Process (ASP) is arguably the most popular bioprocess utilized in the treatment of polluted water. The ASP is described by means of a nonlinear model and results on a Two-Input Two-Output multivariable system. In this paper a discrete time digital control is proposed where the design of a decentralized controller is faced. Local controllers are given the form of a Two-Degree-of-Freedom PI controller tuned using the data-driven Virtual-Reference Feedback tuning approach.

Keywords: ASP, Process Control, PID, Data-Driven Control, VRFT.

1 Introduction

Water pollution represents one of the most serious environmental problems due to the discharge of nutrients into receiving waters. Hence, stricter standards for the operation of wastewater treatment plants (WWTPs) have been imposed by authorities. In order to meet these standards, improved control of WWTPs is needed. Wastewater treatment control has begun a gradual progress towards the use of more advanced technology, in the face of more stringent modern water quality standards. Several approaches have been reported in the literature that attempt to control the WWTPs process. Among others, the Activated Sludge Process (ASP) is arguably the most popular bioprocess utilized in the treatment of polluted water, using microorganisms present within the treatment plant in the biological oxidation of the wastewater. The simplified but still realistic and highly non-linear four-state multivariable model considered here is the ASP as presented in [1].

The ASP is described by means of a nonlinear model and results on a Two-Input Two-Output (TITO) multivariable system. In this paper a discrete time digital control is proposed where the design of a decentralized controller is faced. The paper designs the local controllers as discrete time Proportional-Integral (PI) controllers. The controllers are synthesized using the Virtual Reference Feedback Tuning, which is a model-free based approach where just purely data taken from the system is considered, therefore there is no need for a mathematical model of the system. On the basis of these data the discrete time PI controllers are tuned.

2 The two-degree-of-freedom Virtual Reference Feedback Tuning

The Virtual Reference Feedback Tuning (VRFT) is a one-shot data-based method for the design of feedback controllers. The original idea was presented in [2], and then formalized by

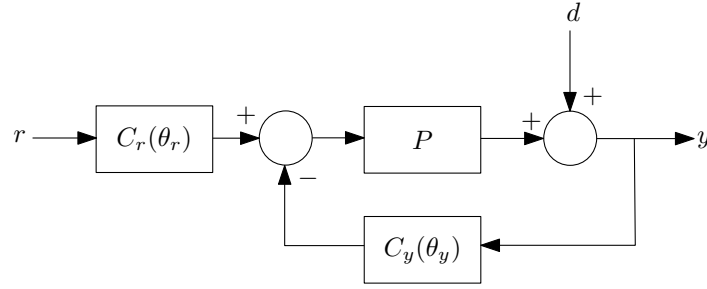


Figure 1: Two degrees of freedom structure

Lecchini, Campi and Savaresi (see [3–5]). In this section, an outline of the two-degree-of-freedom case is presented. The design methodology is presented in [5], the control structure is presented in Fig. 1. The objective of this method is to minimize the criterion in (1).

$$J_{MR}(\theta_r, \theta_y) = \|(\Psi_M(z; [\theta_r, \theta_y]) - M(z))W_M(z)\|_2^2 + \|(\Psi_S(z; \theta_y) - S(z))W_s(z)\|_2^2 \quad (1)$$

with

$$\Psi_M(z; [\theta_r, \theta_y]) = \frac{P(z)C_r(z; \theta_r)}{1 + P(z)C_y(z; \theta_y)} \quad \Psi_S(z; \theta_y) = \frac{1}{1 + P(z)C_y(z; \theta_y)} \quad (2)$$

and $M(z)$ being the target input-to-output transfer function and $S(z)$ the target sensitivity function. In the VRFT framework a plant model is not available, and is not intended to find one. Instead, a batch of input/output data is taken from an experiment on the plant (namely input $u(t)$ and output $y(t)$). So, in order to find the parameters of the controllers (θ_r and θ_y) the signals $\bar{r}(t)$, $\bar{d}(t)$ and $\bar{y}(t)$ are defined. These signals are called “virtual” because they are not really measured, but constructed from the input/output data available and the desired closed-loop relations as follows:

- $\bar{r}(t)$ is the virtual reference, so that $y(t) = M(z)\bar{r}(t)$
- $\bar{d}(t)$ is the virtual perturbation, so that $y(t) + \bar{d}(t) = S(z)\bar{d}(t)$
- $\bar{y}(t)$ is the virtual-perturbed output of the plant, so that $\bar{y}(t) = y(t) + \bar{d}(t)$

These signals are the ones that would be found if $u(t)$ and $y(t)$ had been measured in closed-loop and if the closed-loop dynamics were given by M and S i.e., if the perfect controllers were set in the loop. On the basis of these “virtual” signals the controller’s parameters are found by minimizing the following alternative identification cost function:

$$J_{VR}^N(\theta_r, \theta_y) = \frac{1}{N} \sum_{t=1}^N [\Gamma_M(t; [\theta_r, \theta_y])]^2 + \frac{1}{N} \sum_{t=1}^N [\Gamma_S(t; [\theta_r, \theta_y])]^2$$

where

$$\Gamma_M(t; [\theta_r, \theta_y]) = L_M(z)(u(t) - C_r(z; \theta_r)\bar{r}(t) + C_y(z; \theta_y)y(t)) \quad (3)$$

$$\Gamma_S(t; [\theta_r, \theta_y]) = L_S(z)(u(t) + C_y(z; \theta_y)\bar{y}(t)) \quad (4)$$

and $L_M(z)$ and $L_S(z)$ are appropriate filters to be chosen so (3) becomes an approximation to (1). If the controllers are linear in the parameter ($C_r(z; \theta_r) = \beta_r(z)^T \theta_r$ and $C_y(z; \theta_y) = \beta_y(z)^T \theta_y$) the cost criterion (3) becomes a standard quadratic optimization problem. In [5] the authors use the

concept of “ideal controller” to derive the structure of filters L_M and L_S . The *ideal controllers* C_{r0} and C_{y0} are the ones that, if used in the control loop, would solve (1) exactly, that is

$$C_{y0} = \frac{1-S}{SP} \quad C_{r0} = \frac{M}{SP} \quad (5)$$

When comparing (1) and (3) using the Parseval Theorem the expression of the filters L_M and L_S that must make the identification problem (3) match the control problem (1) are found to be:

$$|L_M|^2 = |M|^2|S|^2|W_M|^2 \frac{1}{\Phi_u} \quad |L_S|^2 = |S-1|^2|S|^2|W_S|^2 \frac{1}{\Phi_u} \quad (6)$$

3 2-DoF PI structure for the VRFT

In order to apply the VRFT framework to the Activated Sludge Process, the structure of the controllers has to be decided before the optimization is carried out. In [6], a decentralized PI structure is used in the same plant with good results for both, reference tracking and disturbance rejection. In this paper, a discretized version of the PI controller is used as the chosen structure for the VRFT controllers. Using a two-degree-of-freedom PI as in Fig. 1, the continuous time version of the controller is:

$$C_r(s) = K_c \left(\beta + \frac{1}{T_i s} \right) \quad C_y(s) = K_c \left(1 + \frac{1}{T_i s} \right) \quad (7)$$

When applying the bilinear transformation $s = \frac{2}{T_s} \frac{z-1}{z+1}$, the controllers are

$$C_r(z) = \frac{K_c \left(\beta + \frac{T_s}{2T_i} \right) + K_c \left(\frac{T_s}{2T_i} - \beta \right) z^{-1}}{1 - z^{-1}} \quad (8)$$

$$C_y(z) = \frac{K_c \left(1 + \frac{T_s}{2T_i} \right) + K_c \left(\frac{T_s}{2T_i} - 1 \right) z^{-1}}{1 - z^{-1}} \quad (9)$$

From the VRFT point of view, (8) and (9) can be seen simply as linear-in-the-parameters controllers with two parameters as follows:

$$C_r(z) = \frac{\alpha_1 + \alpha_2 z^{-1}}{1 - z^{-1}} \quad C_y(z) = \frac{\gamma_1 + \gamma_2 z^{-1}}{1 - z^{-1}} \quad (10)$$

Since the continuous time controllers (7) have three adjustable parameters (K_c , T_i , β), one of the four parameters in its discrete time equivalent (10) should depend of the other three. From (8) to (10) it can be found that

$$\alpha_1 + \alpha_2 = \gamma_1 + \gamma_2 = \frac{K_c T_s}{T_i} \quad (11)$$

and we have that

$$\gamma_2 = \alpha_1 + \alpha_2 - \gamma_1 \quad (12)$$

Then the discrete time controllers (10) are now

$$C_r(z) = \frac{\alpha_1 + \alpha_2 z^{-1}}{1 - z^{-1}} \quad C_y(z) = \frac{\gamma_1 + (\alpha_1 + \alpha_2 - \gamma_1) z^{-1}}{1 - z^{-1}} \quad (13)$$

Once the parameters of the controllers (13) ($\alpha_1, \alpha_2, \gamma_1$) are found, one can recuperate the PI (7) parameters using:

$$K_c = \gamma_1 - \frac{1}{2}(\alpha_1 + \alpha_2) \quad (14)$$

$$T_i = T_s \frac{\gamma_1 - \frac{1}{2}(\alpha_1 + \alpha_2)}{\alpha_1 + \alpha_2} \quad (15)$$

$$\beta = \frac{\alpha_1 - \frac{1}{2}(\alpha_1 + \alpha_2)}{\gamma_1 - \frac{1}{2}(\alpha_1 + \alpha_2)} \quad (16)$$

4 Activated Sludge Process (ASP) Description

The mathematical model considered in this paper is given in [1]. The ASP process comprises an aerator tank where microorganisms act on organic matter by biodegradation, and a settler where the solids are separated from the wastewater and recycled to the aerator. The layout is shown in Fig. 2. The component balance for the substrate, biomass, recycled biomass and dissolved oxygen provide the following set of non-linear differential equations:

$$\frac{dX(t)}{dt} = \mu(t)X(t) - D(t)(1+r)X(t) - rD(t)X_r(t) \quad (17)$$

$$\frac{dS(t)}{dt} = -\frac{\mu(t)}{Y}X(t) - D(t)(1+r)S(t) + D(t)S_{in} \quad (18)$$

$$\begin{aligned} \frac{dDO(t)}{dt} &= -\frac{K_o\mu(t)}{Y}X(t) - D(t)(1+r)DO(t) \\ &+ K_{La}(DO_s - DO(t)) + DO(t)DO_{in} \end{aligned} \quad (19)$$

$$\frac{dX_r(t)}{dt} = D(t)(1+r)X(t) - D(t)(\beta+r)X_r(t) \quad (20)$$

$$\mu(t) = \mu_{max} \frac{S(t)}{k_S + S(t)} \frac{DO(t)}{k_{DO} + DO(t)} \quad (21)$$

where $X(t)$ - biomass, $S(t)$ - substrate, $DO(t)$ - dissolved oxygen, DO_s - maximum dissolved oxygen, $X_r(t)$ - recycled biomass, $D(t)$ - dilution rate, S_{in} and DO_{in} - substrate and dissolved oxygen concentrations in the influent, Y - biomass yield factor, μ - biomass growth rate, μ_{max} - maximum specific growth rate, k_S and k_{DO} - saturation constants, $K_{La} = \alpha W$ - oxygen mass transfer coefficient, α - oxygen transfer rate, W - aeration rate, K_o - model constant, r and β - ratio of recycled and waste flow to the influent. The influent concentrations are set to $S_{in} = 200$ mg/l and $DO_{in} = 0.5$ mg/l. With respect to the control problem definition, the waste water treatment process is considered under the assumption that the dissolved oxygen, $DO(t)$, and substrate, $X(t)$, are the controlled outputs of the plant, whereas the dilution rate, $D(t)$, and aeration rate $W(t)$ are the two manipulated variables. The initial conditions and kinetic parameters are taken as in [1] and [6].

5 Discrete-time VRFT tuned PI controller applied to the ASP

Using the non-linear model presented in (17) to (21), the data in Fig. 3(a) and Fig. 3(b) was collected. Using only this data without the information of the non-linear model, or any linear approximation, the parameters of the controller are calculated according to Section 2 using the PI structure specified in Section 3. The VRFT controllers are tested against two decentralized

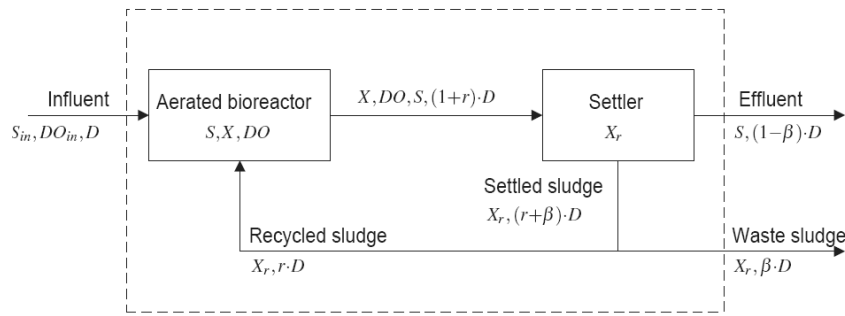
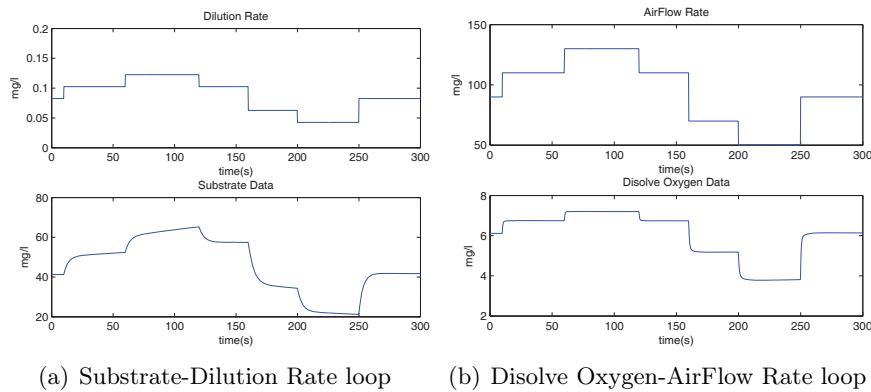


Figure 2: Activated Sludge Process layout. Taken from [7]



(a) Substrate-Dilution Rate loop (b) Dissolve Oxygen-AirFlow Rate loop

Figure 3: Data used to find the VRFT controller

PI controllers which parameters are computed using IMC [8], based on considering a First-Order (FO) controlled-process given by

$$P(s) = \frac{K_p}{T_s + 1} \quad (22)$$

The identified models obey to $K_1 = 437.1$ and $T_1 = 2.7h$ for the first loop and $K_2 = 0.03$ and $T_2 = 0.51h$ for the second loop. These values are as in [6]. The controllers were discretized using the bilinear transformation (sampling time of 15min). In the case of the VRFT, the controllers are directly found in discrete time. The closed-loop specifications are given in terms of the desired time constants of the controlled system for each loop. The desired time constant of the Substrate-Dilution Rate loop is approximately $T_1 = 2.7h$ while the Dissolved Oxygen-Airflow Rate loop constant time is approximately $T_2 = 0.51h$. The closed-loop constant time for each variable is giving in terms of the dimensionless variables τ_{c1} and τ_{c2} via $T_{c1} = \tau_{c1}T_1$ and $T_{c2} = \tau_{c2}T_2$. If, for example, $\tau_{c1} < 1$, the Substrate-Dilution Rate closed-loop is expected to be faster than in open loop. It is worth to note that for the VRFT tuning it is possible to specify a different time constant for the disturbance attenuation transfer function (in fact $S(z)$). In this case, the VRFT tuned PI controllers are found with a time constant for the corresponding $S(z)$ transfer function that is half the one specified for the reference to output relation by using τ_c (therefore $\tau_c/2$). If $\tau_{c1} = \tau_{c2} = 1$, the resulting controller parameters are $K_{c1} = 0.0023$, $K_{c2} = 33.33$, $T_{i1} = 2.7h$ and $T_{i2} = 0.51h$ for the IMC and $K_{c1} = 0.0042$, $K_{c2} = 25.41$, $T_{i1} = 3.43h$, $T_{i2} = 0.42h$, $\beta_1 = 0.6602$ and $\beta_2 = 0.74$ for the VRFT. The response to a change in the set points of both loops is giving in Fig. 4. The IAE value represents the Integrated Absolute Value of the Error. As it can be seen, the responses for the VRFT provide smaller IAE as well as less demanding control actuation, computed here as the Total Variation (TV) or the sum of the control movements from

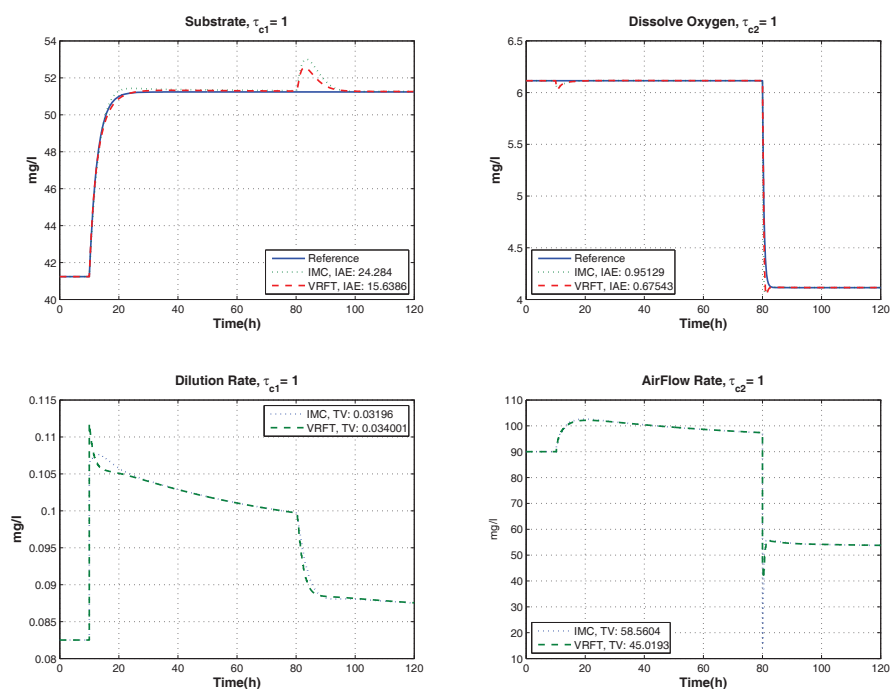


Figure 4: VRFT and IMC controllers responses to set-points step changes, $\tau_{c1} = \tau_{c2} = 1$

one sampling time to the other.

In case the closed-loop bandwidth is increased and we set $\tau_{c1} = \tau_{c2} = 0.5$ the resulting controller parameters are $K_{c1} = 0.0046$, $K_{c2} = 66.67$, $T_{i1} = 2.7h$ and $T_{i2} = 0.51h$ for the IMC and $K_{c1} = 0.0065$, $K_{c2} = 33.38$, $T_{i1} = 2.68h$, $T_{i2} = 0.35h$, $\beta_1 = 0.7752$ and $\beta_2 = 1.03$ for the VRFT. In this case the IMC controller presents and undesirable oscillating behavior in the Dissolved Oxygen loop, as shown in Fig. 5. The output of the controllers was saturated to 0, in case it went below this value. Also the control effort of the second controller is quite lower in the case of the VRFT with a performance nearly 50% better. Also a simulation was carried out for a disturbance in the inflow concentration. The results are depicted in Fig. 6. Again the results are quite similar, and the multivariable characteristic is tackled in a satisfactory way. As it can be seen, this data driven methodology is suitable for the control of the ASP process and it allows to skip the modeling step in order to find a good controller.

6 Conclusions

This paper has presented the application of a purely data based approach for tuning of discrete time PI controllers. The main advantage of the proposed method is that it does not rely on the usual linear model approximation of the system to be controlled. Just an experiment that provides input-output data from the system is needed. The performance of the tuning approach has been tested on a non-linear multivariable system and performance compared with that of the well known IMC method. As designed performance is more demanding, the resulting control system exhibits better results than its IMC counterpart.

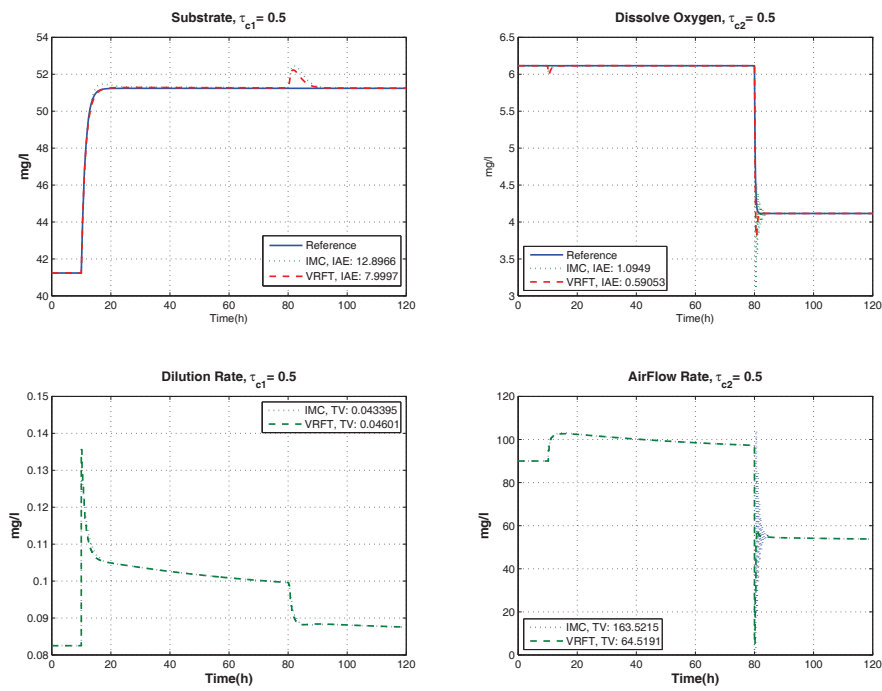


Figure 5: VRFT and IMC controllers responses to set-points step changes, $\tau_{c1} = \tau_{c2} = 0.5$

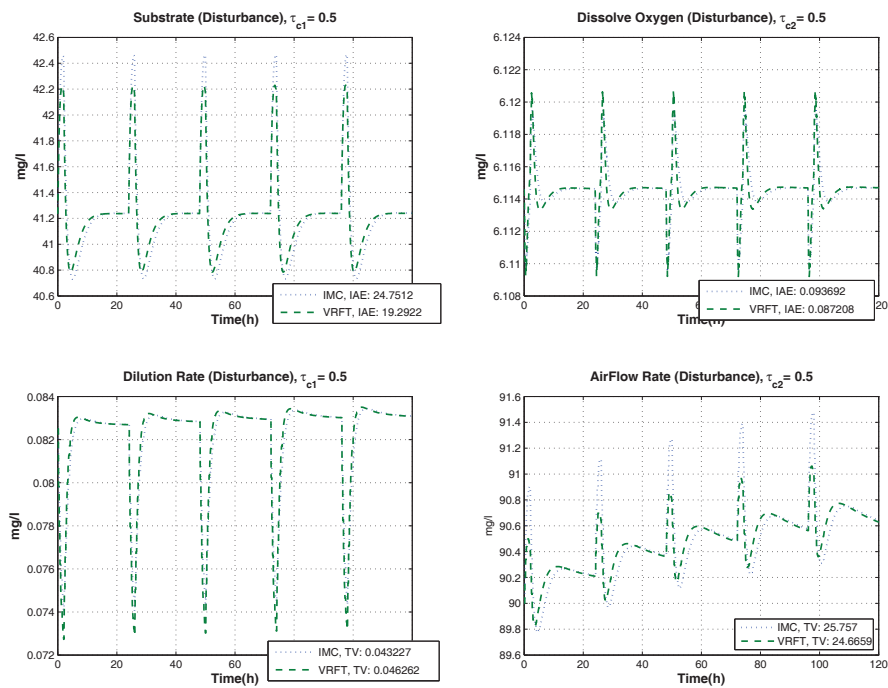


Figure 6: Responses changing the specification for the VRFT and comparison with the IMC controller, $\tau_{c1} = \tau_{c2} = 0.5$

Acknowledgment

This work has received financial support from the AEI-PCI program A/025100/09 and from the Spanish CICYT program under grant DPI2007-63356. Research work by J.D. Rojas has received financial support from the Universitat Autònoma de Barcelona. Support from the Universidad de Costa Rica is greatly appreciated.

Bibliography

- [1] F. Nejari, A. Benhammou, B. Dahhou, and G. Roux, “Non-linear multivariable adaptive control of an activated sludge wastewater treatment process,” *Int. J. Adapt. Control Signal Process.*, pp. 347–365, 1999.
- [2] G. Guardabassi and S. Savaresi, “Virtual reference direct design method: an off-line approach to data-based control system design,” *Automatic Control, IEEE Transactions on*, vol. 45, pp. 954–959, May 2000.
- [3] M. C. Campi, A. Lecchini, and S. M. Savaresi, “Virtual reference feedback tuning: a direct method for the design of feedback controllers,” *Automatica*, vol. 38, no. 8, pp. 1337 – 1346, 2002.
- [4] A. Lecchini, M. Campi, and S. Savaresi, “Sensitivity shaping via virtual reference feedback tuning,” in *Decision and Control, 2001. Proceedings of the 40th IEEE Conference on*, vol. 1, pp. 750–755 vol.1, 2001.
- [5] A. Lecchini, M. Campi, and S. Savaresi, “Virtual reference feedback tuning for two degree of freedom controllers,” *International Journal of Adaptive control and Signal Processing*, vol. 16, no. 5, pp. 355–371, 2002.
- [6] R. Vilanova, R. Katebi, and V. Alfaro, “Multi-loop PI-based control strategies for the Activated Sludge Process,” in *Emerging Technologies and Factory Automation, 2009. ETFA 2009. IEEE International Conference on*, September 2009.
- [7] S. Caraman, M. Sbarciog, and B. Marian, “Predictive Control of a Wastewater Treatment Process,” *International Journal of Computers, Communications & Control*, vol. 2, no. 2, pp. 132–142, 2007.
- [8] M. Morari and E. Zafirou, *Robust Process Control*. Prentice-Hall International, 1989.