Model-based methodology for the design of optimal control strategies in MBR plants
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ABSTRACT
This paper proposes a model-based methodology that allows synthesising the most appropriate strategies for optimising the operation of wastewater treatment plants (WWTPs). The methodology is applied with the aim of maximising the nitrogen removal in membrane bioreactors (MBRs). The proposed procedure is based on a systematic approach composed of four steps. First, a sensitivity analysis of the input variables is carried out in order to obtain a first assessment of the potential for operational improvements. Then, the optimum input variable values are calculated by a model-based optimisation algorithm that minimises a cost function associated with the effluent total nitrogen at different temperatures. Then, the optimum operational strategies are identified. Finally, these operational strategies form the conceptual knowledge base for designing automatic control laws. The obtained optimal control strategies have shown a significant improvement in performance in comparison with fixed operation for the studied case, reducing the total nitrogen by 40%.

Key words | MBR, model-based, optimisation, operation, WWTP

INTRODUCTION
Nowadays, the use of mathematical models and simulations of wastewater treatment plants has become very important for optimising their design and operation. In the last decades, several models that dynamically describe the biochemical transformations taking place in the biological processes have been developed (Henze et al. 2000). One of the main advantages of mathematical modelling and computer simulation is the capacity to analyse many different scenarios with very little effort. This is a critical property for optimisation algorithms, since a lot of simulations need to be carried out in order to locate the optimal solution and, in real life, this would be unfeasible in terms of time and budget.

Furthermore, wastewater treatment plants (WWTPs) are commonly designed for critical conditions but they are working at under-loaded conditions most of the time, offering a great opportunity for optimising their operation. All the possible combinations of the input variables (wastage flow, dissolved oxygen set-point, etc.) define the feasible operating space of the plant. However, some of the points within the feasible region do not comply with the process requirements, so these operational points should be avoided. Hence, the allowable operating zone is a subspace of the feasible operating space where those points are not included. Although all the points within the allowable operating zone are suitable for operating
the plant, each of them can produce different outputs in terms of consumed energy or effluent quality. Hereby, by properly selecting the operating point of the membrane bioreactor (MBR) plant, its performance can be optimised. However, since the state of the MBR plant is constantly fluctuating due to influent or temperature disturbances, the optimum operating point is also permanently varying. Thus, instead of using a fixed optimal operating point, a set of generic control laws for constantly optimising the plant performance is proposed. These control laws will be synthesised using several model-based optimisations at different plant temperatures.

The optimal operation of the conventional activated sludge (CAS) plant has been widely studied (Galarza et al. 2001), but the implementation of the new MBR technology has introduced several differences with the CAS technology. On the one hand, the MBR can be operated at higher total suspended solids (TSS) concentrations, which leads to a better biological performance and filtration. On the other hand, the distribution of the solids is also very different, since in the MBR technology there is a significant gradient of solids between the MBR and the rest of the tanks (Beltrán et al. 2009). Additionally, MBRs are operated at a constant air scour flow rate, which is normally not lowered because of membrane fouling potential (Judd & Judd 2011). All these factors can affect the performance of the process and, therefore, the operational strategies that are commonly applied for CAS plants should be revised when membrane reactors are incorporated. Nopens et al. (2007) studied the optimisation of the biological performance of a side stream MBR. Verrecht et al. (2010) proposed a model-based optimisation of a small-scale decentralised MBR for enhancing energy savings and biological efficiency. Lim et al. (2011) optimised the operational conditions of an MBR for maximising the chemical oxygen demand (COD) and nitrogen removal. Dalmau et al. (2015) carried a model-based study of the integrated operation of a nutrient removal pilot scale MBR. Mannina & Cosenza (2013) present an integrated mathematical model for minimising energy costs. Gabarrón et al. (2015) propose a mechanistic model for reducing the aeration energy costs. However, all these studies share a common limitation, they lack control laws for optimising the plant performance despite operational disturbances.

Thus, this paper proposes a model-based systematic methodology for synthesising control strategies for optimising the operation of MBRs. This methodology has been applied for maximising nitrogen removal in MBR plants.

## METHODS

### Model-based construction of control laws

The proposed procedure for the synthesis of operational strategies and controllers in MBR plants is based on a systematic approach composed of four consecutive steps:

#### Simulation-based exploration of the operating scenario

This first step of the procedure aims at assessing the effect of manipulating the input variables, by changing the influent load and the temperature, in the final process performance. Simulation results can be normally condensed in sensitivity plots or nomograms to facilitate their interpretation. If simulations suggest that an adequate manipulation of input variables offers a significant potential for improving process efficiency at different scenarios, the next step can be launched.

#### Model-based calculation of the optimum operating points

This second step proposes the application of model-based mathematical optimisation algorithms for the automatic calculation of the most appropriate sets of manipulated variables in all the scenarios under study. Each optimisation problem requires the definition of the degrees of freedom (typically the free manipulated variables), the restrictions (requirements or boundaries) and the cost function (normally associated with effluent quality or economical costs). A complete description of the optimisation algorithm can be found in Rivas et al. (2008). The result of this step is the set of optimum operational points for the predefined operational objectives and restrictions.

#### Identification of the optimum operational strategies

For this purpose, the trajectories of both the optimum operational points and the state of the process should be related, to identify the criteria for optimising process performance under changing scenarios. These criteria can be qualitative or quantitative, and they are normally associated with rules or properties that are met by most of the optimum points under changing conditions and, consequently, they are not significantly affected by process perturbations.
Design and model-based validation of the automatic controllers

This final step consists of transforming (when possible) these optimum operational rules to automatic control loops, capable of selecting the most appropriate value of the manipulated variables at each moment using the information provided by the available measurement data. A final model-based validation should be carried out in order to confirm or refute the initial expectations in process improvements.

Description of the case study: MBR for N removal

The proposed model-based procedure for designing operational strategies has been applied to construct and validate the most appropriate automatic controllers for maximising N removal in MBR reactors. The following virtual plant will be used for this paper.

Figure 1 shows the predenitrification-nitrification MBR plant layout used as case-study, which is one of the most common configurations for nitrogen removal. The MBR plant is composed of five tanks: the first two tanks are in anoxic conditions (without external aeration) for denitrification and the other three will remain in aerobic conditions for the nitrification process. Nitrates produced in the aerobic tanks are sent to the anoxic tanks by the recirculation flow QR, and the total solids of the system are controlled by the wastage flow Qw. The final MBR tank is operated at a minimum constant aeration flow, calculated to prevent an unsuitable fouling of the membrane using the membrane manufacturer recommendations. ASM2d has been used for the biological transformations at the reactors. Since this paper is focused in the biological performance of the MBRs the membrane fouling has not been taken into account. Characteristics of influent load are presented in Table 1.

The minimum volume of the MBR tank has been estimated using commercial information from membrane manufacturers. The optimum dimensions of the other five plant reactors have been automatically calculated looking for the minimum volume of the plant that is able to fulfil the required effluent requirements at critical conditions (15 °C) (Rivas et al. 2008). The restriction of assuming similar volume for each reactor has not a significant effect in the results. Other possible plant layouts, like the BSM-MBR (Maere et al. 2011), have also been analysed but they don’t show a significant improvement in process performance.

RESULTS

Simulation-based exploration of the operating scenario

A simulation-based analysis has been carried out with the aim of assessing the effect of the main input variables (recirculation ratio and dissolved oxygen) to N-NH4 effluent concentration and total effluent nitrogen concentration (considered simply as the sum of N-NH4 and N-NO3 concentrations). The recirculation ratio (QR / QINF) has been changed between 1 and 8 (8 being the upper limit of the pump) with a 0.1 step size and the dissolved oxygen concentration in the aerobic tanks has been operated in the 0.0–2.0 g O2 m⁻³ range with a 0.1 g O2 m⁻³ step size. This way the potential for improvement of the plant performance

Table 1 | Characterisation of the influent wastewater

<table>
<thead>
<tr>
<th>QINF (m³ d⁻¹)</th>
<th>SΝH₄ (g N m⁻³)</th>
<th>X₄ (g COD·m⁻³)</th>
<th>X₇SS (g TSS·m⁻³)</th>
</tr>
</thead>
<tbody>
<tr>
<td>19,400</td>
<td>24.2</td>
<td>31.3</td>
<td></td>
</tr>
<tr>
<td>27.3</td>
<td>7.0</td>
<td></td>
<td></td>
</tr>
<tr>
<td>68.8</td>
<td>56.2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>4.6</td>
<td>222.9</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
can be assessed. Moreover, it is very important to study the shape of the feasible operating space so that very sensitive optimum zones can be avoided to assure the stability of the WWTP.

Figure 2 shows the total nitrogen concentration isolines in continued lines while the ammonium concentration isolines are represented by dotted lines; both variables are shown in gN/m³. Two different temperatures are presented: 15 °C and 21 °C.

First of all, it can be observed that the temperature does not significantly affect the qualitative effect of the operational variables. Besides, the ammonium concentration decreases when the oxygen in the aerobic tanks R4 and R5 is increased, as expected. Moreover, the ammonium can be reduced by the increase in the recirculation rate. Furthermore, the total nitrogen first decreases with the oxygen and then starts to increase, due to the equilibrium between the eliminated ammonium and the produced nitrates. Finally, the total nitrogen at first decreases when the recirculation rate is increased, but there is a limit from which the total nitrogen begins to increase. This behaviour is mainly caused by the equilibrium between increasing solids concentrations in the reactors and sending higher oxygen to the anoxic tanks.

Finally, the design point (DP) and the optimal point (OP) have been represented. The DP is the value of the input variables at the design temperature (15 °C), while the OP is the operational point that, maintaining the volume distributions, minimises the total nitrogen concentration of the effluent and meets the ammonium concentration constraint at the temperatures under study (15 °C and 21 °C in the examples). The red arrows represent how the optimal operational point has shifted with the temperature changes. It can be seen that, in both cases, the lowest effluent total nitrogen is achieved at low oxygen set points and high recirculation ratios. However, the restriction of the effluent ammonia (1 gN/m³) prevents operating in that zone.

It is interesting to note that the total amount of nitrogen could be theoretically reduced by 22% and 51% at 15 °C and 21 °C respectively. Hence, the performance of the plant has the potential to be greatly enhanced by an optimal operational strategy. Moreover, it can be seen that the operating space is smooth and continuous. Thereby, small disturbances will have little impact in the OP. This facilitates the design of optimal controllers.

Once it has been seen that there is room for optimising the plant operation, the next step is to calculate the optimal operational points.

Model-based calculation of the optimum operating points

The main purpose of this section is to optimise the input variables of the plant at different temperatures, and based on these optimisations generic control laws will be synthesised. For this purpose, the first step is the calculation of the optimal values of the operational variables at different temperatures (between 15 °C and 23 °C) that minimise the total effluent nitrogen satisfying the constraints. Table 3 summarises the optimisation problem solved at each temperature.
Previous optimisations (not shown) have demonstrated that the evolution of optimum dissolved oxygen at both reactors R4 and R5 is nearly similar. Therefore, for simplicity purposes, R4 and R5 are forced to have the same dissolved oxygen concentration. This additional constraint has reduced the degrees of freedom of the problem, facilitating the further selection of the operational strategy.

The results of the optimisation problem at different temperatures are shown in Figure 3.

The upper figure (Figure 3(a)) shows the evolution of the optimal input variables (dissolved oxygen and recirculation ratio) while the lower figure (Figure 3(b)) presents the effluent concentration (nitrates and ammonium) obtained by the corresponding operating point at each temperature.

It has been commonly observed that, for under loaded conditions, the reduction of dissolved oxygen in the aerated reactors reduces significantly the total effluent nitrogen due to the enhanced denitrification (Dalmau et al. 2014). In the MBR plant under study, there are two oxygen related input variables: DO_{R3} and DO_{R4} – DO_{R5}. The reduction of DO_{R3} will lead to a predenitrification-nitrification configuration (DN), while decreasing DO_{R4} – DO_{R5} will lead to a two in-series DN configuration (DNDN). Optimisation results have clearly shown that this second option is most appropriate for optimising the plant under higher temperatures. The first anoxic zone (R1 and R2) is in charge of reducing the recirculated nitrates using influent COD, and the second anoxic zone (R4 and R5) carries out a second denitrification process reducing the nitrates produced in R3. Figure 4 shows the resulting evolution of the optimum plant layout when the wastewater passes from cold temperatures to warm temperatures.

Figure 3(a) also shows that the optimum recirculation ratio increases with temperature until its maximum value is reached. This rise produces higher nitrate recirculation flux (which compensates for the reduction in nitrate concentration), higher suspended solids in the plant (because the maximum concentration of 10,000 g TSS m$^{-3}$ in the MBR tanks has been maintained) and higher introduction of dissolved oxygen in the anoxic zones. The optimal recirculation flow will result from a balance between them and, since the two first positive effects are reinforced with the temperature, the optimum QR value increases along with it.

Figure 3(b) shows the minimal total effluent nitrogen that can be reached by the plant at different temperatures between 13°C and 23°C. It can be seen that the optimum ammonium concentration always remains constant at its maximum value of 1.0 g N m$^{-3}$, while the nitrates are progressively reduced up to 2.1 g N m$^{-3}$ making possible a total effluent nitrogen concentration of 3.1 g N m$^{-3}$.

These results confirm the great potential for optimising the operation of the plant to achieve the minimum effluent total nitrogen.

### Identification of the optimum operational strategies

Once the evolution of the optimal values of input variables has been analysed, the next crucial point is the synthesis of an optimum (or sub-optimum) set of operational strategies that could be practically implemented (manually or automatically) in the plants. At this point, it is very important to remark that these strategies should be clear, realistic and based on measurable and reliable information. Model simulation results have shown their essential role in analysing the process dynamics and the manipulation effects.

#### Table 3 | Definition of the optimisation problem for each temperature

<table>
<thead>
<tr>
<th>Objective function</th>
<th>Minimal total nitrogen in the effluent (NH$<em>4$$</em>{\text{eff}}$ + NO$<em>3$$</em>{\text{eff}}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Variables</td>
<td>DO$<em>{R3}$; DO$</em>{R4}$; DO$<em>{R5}$; Q$</em>{R}$; Q$_{w}$</td>
</tr>
<tr>
<td>Constraints</td>
<td>DO$<em>{R3}$ ≤ 2 g O$<em>2$ m$^{-3}$ DO$</em>{R4}$ ≤ 2 g O$<em>2$ m$^{-3}$ DO$</em>{R5}$ ≤ 2 g O$<em>2$ m$^{-3}$ DO$</em>{R4}$ = DO$</em>{R5}$ Q$<em>R$ &lt; = 8 · Q$</em>{\text{INF}}$ TSS$_{\text{MBR}}$ ≤ 10 g L$^{-1}$ NH$<em>4$$</em>{\text{eff}}$ ≤ 1 g N m$^{-3}$ NO$<em>3$$</em>{\text{eff}}$ ≤ 8 g N m$^{-3}$</td>
</tr>
</tbody>
</table>

![Figure 3](https://example.com/figure3.png)  
**Figure 3** | Optimal variables and evolution of the minimal total nitrogen.
however, the resulting operational rules should be generic
and independent of the specific model results or predictions.

From the analysis of the results obtained in the model-
based optimisation, several rules for optimum operation of
this kind of MBR plant with N removal can be extracted:

- Dissolved oxygen in R4-R5 should be adjusted to strictly
accomplish the effluent ammonium requirements. This
strategy moves the plant to a DNDN plant layout that
enhances denitrification and reduces aeration costs.
Additionally, it maintains the nitrification activity in the
membrane reactor, avoiding over oxygenation in the
recirculation ratio to the anoxic zones. It is important
to remember that aeration in the membrane reactor
cannot be reduced to prevent membrane fouling.

- The recirculation ratio should compensate for the vari-
ations in effluent nitrate concentration. For increasing
temperatures (or reducing loads), the recirculation
flow should be progressively increased in order to supply the
nitrates required for denitrification in R1 and R2.

- The sludge wastage rate should be selected with the aim
of maintaining (in the long term) the required solids con-
centration in the membrane reactor. It is interesting to
note that this concentration is also perturbed by the vari-
ations in the recirculation ratio.

The next step of the procedure is to design the controllers to apply the synthesised optimal control laws.

Design and model-based validation of the automatic
collectors

The goal of this section is to design an automatic controller
so that a plant can be optimally operated at any temperature
despite disturbances. First, the design of the optimum control-
trollers is explained and then, different tests of the
controllers have been carried out.

As it has been shown in Table 3, five input variables are considered. However, from the optimisation results, it can be
seen that the optimal DO concentration of the aerobic
tank R3 does not change with the temperature, so this variable
does not need to be controlled. For reasons of simplicity, the DO concentration in the aerobic reactors
R4 and R5 are considered to be the same so they will
share the same controller. It has been checked via optimisations that this hypothesis has very little effect on the total
effluent nitrogen. Consequently, just three controllers will
be analysed: a DO controller C1, a recirculation flow con-
troller C2, and a wastage flow controller C3. Table 4 shows the controlled variables, their set-points, the input
variables and their constraints.

Controller C1 automatically manipulates the common
DO set-point in reactors R4-R5 to strictly maintain the
required ammonium concentration in the effluent (1.0 g N m⁻³ in this example). This control strategy was successfully
validated in full-scale plants (Ayesa et al. 2006) but, as said before, the specific characteristics of MBR plants tends to
move the plant layout to a very efficient DNDN configura-
tion during under loaded conditions.

Controller C2 is a very well-known loop that manip-
ulates the recirculation flow in order to maintain a
minimum (but higher than zero) nitrate concentration at
the end of the anoxic volume. This loop has been succes-
sfully validated at full-scale plants (Ayesa et al. 2006). It
should be noted that a minimal recirculation rate of 3.4 is
used to maintain a minimal biomass concentration.

Controller C3 regulates the long-term amount of
solids in the system to guarantee appropriate conditions
(suspended solids) for the membrane filtration. It is interest-
ing to remark that this loop should have a slow dynamic
decoupled from possible fast perturbations in the solids dis-
tribution among tanks.

Three incremental PI controllers (Åström & Hägglund
1995) have been programmed and tuned in order to carry
out a first assessment of the control strategy under dynamic
conditions. For testing these controllers, a one year influ-

<table>
<thead>
<tr>
<th>Control name</th>
<th>C1</th>
<th>C2</th>
<th>C3</th>
</tr>
</thead>
<tbody>
<tr>
<td>Controlled variable</td>
<td>NH₄,eff</td>
<td>NO₃,R2</td>
<td>TSSMBR</td>
</tr>
<tr>
<td>Set point</td>
<td>1 g N m⁻³</td>
<td>0.5 g N m⁻³</td>
<td>10,000 g TSS m⁻³</td>
</tr>
<tr>
<td>Control action</td>
<td>DO₄₅₅</td>
<td>QR</td>
<td>Qw</td>
</tr>
<tr>
<td>Minimum</td>
<td>0 g O₂ m⁻³</td>
<td>3.4 QINF</td>
<td>0 m² s⁻¹</td>
</tr>
<tr>
<td>Maximum</td>
<td>2 g O₂ m⁻³</td>
<td>8 QINF</td>
<td>-</td>
</tr>
</tbody>
</table>
with variations in the temperature based on the BSM1_LT (Rosen et al. 2004) has been used.

Table 5 shows the average results for different control strategies that combine the simultaneous switching on of different control loops. Strategy A consist of operating the plant at the fixed operational point selected for design at critical conditions (DP in Figure 2). Strategies B, C and D show the effect of incorporating different loops and, finally, Strategy E simultaneously combines the three loops. It can be clearly seen that C1, the dissolved oxygen control, is the most important controller in the plant since its activation is the crucial factor for decreasing the total nitrogen in the effluent.

Figure 5 shows the evolution of the 24 h average ammonium and nitrate concentrations throughout the dynamic simulation of control strategies A and D. The dark grey line is associated with the closed loop strategy D while the light grey line represents strategy A. The ammonia and nitrate concentrations are shown in the upper half of the graph, and the oxygen set point is represented in the lower half. It can be clearly seen how the denitrification of the controlled plant can be greatly enhanced by the automatic controllers, decreasing the total nitrogen by 40%. It is expected that these very successful results can be improved additionally using more sophisticated controllers (for example, incorporating mobile-averaged windows and predictive actions), but this is part of the current research activity of the research team.

### CONCLUSIONS

The operational optimisation of MBRs is a complex task due to the high amount of variables involved in the biological processes and the continuous disturbances in the influent and temperature. This paper has presented a model-based systematic procedure for analysing the influence of input variables in process performance and the use of automatic model-based optimisation algorithms for designing the most appropriate operational strategies.
The proposed procedure has been applied to synthesise reasonable control strategies for a case study MBR plant. The optimisation results have identified the most suitable strategies for minimising effluent nitrogen in MBRs, which incorporates some remarkable differences from the rules and criteria conventionally used in predenitrification-nitrification plants. Particularly noticeable is the automatic modification of the plant layout from DN to DNDN configuration for optimising nitrogen removal at high temperatures. Finally, the performance of the proposed controllers has been successfully validated by long-term simulations.

Current research activity is focused on using the designed methodology to improve the operational strategies in a full-scale MBR system.

REFERENCES


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