Importance of calibration for mathematical modeling of self-purification of lotic environments

Importância da calibração na modelagem matemática de autodepuração em ambiente lótico

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Abstract: Aim: To demonstrate the importance of calibration in mathematical modeling of self-purification in lotic environments, this study simulated the behavior of various parameters in a river with average annual flows between 4.0 and 32.0 m³.s⁻¹, in a segment downstream from the entry of treated wastewater from a sewage treatment station (average monthly flow of 2.1 m³.s⁻¹).

Methods: Numerical solution by finite difference of the advection-diffusion equation was used to study the dispersion and to quantify and monitor the evolution over time of the parameters DO, BOD₅, P_total, NH₃, NO₃ and the levels of the heavy metals cadmium, chromium, copper, lead and zinc. The longitudinal behavior of the water quality parameters simulated by calibrating the state variables was compared with the behavior of the same parameters simulated via state variables available in the literature. The sensitivity of the state variables was also analyzed.

Results: The calibration process led to good fits between the simulated and actual data for all the parameters analyzed. On the other hand, the comparison of the water quality model using calibrated state variables with the model based on state variables obtained in the literature revealed inconsistencies regarding the parameters DO, P_total, ammonia, nitrate and all the heavy metals.

Conclusions: Considering the wide threshold ranges of the state variables in the literature and the dearth of studies on calibrating the coefficient of decay and quantifying the release of heavy metals by bottom sediment, this study can serve as a base for future investigations in lotic environments with similar hydraulic and water quality characteristics.

Keywords: River; Brazil; water quality; monitoring; model comparison.
Introduction

Socioeconomic development and demographic growth are accompanied by increased demand for water with good quality. Indeed, clean water is a vital resource for the health and development of any community (Bueno et al., 2005).

Multiple uses of surface water and uncontrolled discharge of sanitary wastes cause negative impacts on the quantity and quality of water in urbanized regions (Zhang et al., 2011; McKnight et al., 2012; Sulis & Sechi, 2013; Yenilmez & Aksoy, 2013; Salla et al., 2013a). Among the anthropic actions that affect water bodies, release of inorganic and carbonaceous and nitrogenous organic compounds particularly degrades water quality, since the water has limited capacity for biochemical degradation of supernatant organic matter, so that the dissolved oxygen, among its various functions, is a determining factor for the survival of living organisms in the water. Heavy metals also pose a severe risk of immediate and future damages. These come from varied sources, ranging from natural erosion processes to industrial or mining wastes and pesticides in runoff water from farms. Unlike biodegradable organic compounds, heavy metals are recalcitrant and can accumulate in environmental components and aquatic organisms, causing toxic reactions (Baird, 2002).

Mathematical modeling of water quality in lotic environments is a fundamental tool for planning use and managing water resources on the watershed scale. Many models are available in the literature and the mathematical interpretation of the phenomena and parameters modeled can assume diverse forms and involve varied hypotheses. Therefore, the model needs to be chosen in view of the objectives and variables of interest (Salla et al., 2015).

Regarding the mathematical interpretation of environmental phenomena in lotic environments, the information in the literature on biochemical reaction coefficients, sediment removal coefficients, benthic demands and resurgence is ample. These aspects depend on the hydraulic characteristics of the channel, the prior water quality and the quality of the incoming effluent, among others. Wide variations are presented for various coefficients and demands, which can cause the results obtained by mathematical simulation of water quality to deviate markedly from the real situation (Salla et al., 2014b). With respect to heavy metals, the leading models at present contain mathematical equations reflecting the behavior of metals (often identified under the umbrella term toxic contaminants), but lack threshold indicators of the coefficients included in these models. In other words, they lack support from studies of water courses (be they small, medium or large) for calibration of the coefficients related to the behavior of heavy metals.

In general, mathematical modeling satisfactorily reflects the self-purification process when the coefficients, benthic demands and resurgence features are calibrated based on water quality data monitored along the water course (Salla et al., 2015). Also, the determination in laboratory scale of the benthic demand for dissolved oxygen (DO) is beneficial in mathematical modeling because it reduces the variables to be calibrated, due to the high sensitivity of this factor in the behavior of the parameters DO and biochemical oxygen demand (BOD).

In this context, by using six monitoring points along the Uberabinha River, located in the “Triângulo Mineiro” region of the state of Minas Gerais, Brazil, as the basis for comparison, it was possible to show the importance of calibration in the mathematical modeling of self-purification in a lotic environment, by evaluating the behavior of the longitudinal profiles of the parameters DO, BOD, ammoniac nitrogen (NH$_3$), nitrate (NO$_3$), total phosphorus (P$_{total}$) and the metals cadmium (Cd), chromium (Cr), copper (Cu), lead (Pb) and zinc (Zn).

Materials and Methods

2.1. Study area

The Uberabinha River watershed is located in the western part of the state of Minas Gerais, between the coordinates 18°42’ and 18°54’ south latitude...
and 48°18’ and 48°30’ west longitude (Figure 1). The river’s headwater is located in the municipality of Uberaba, from where it runs about 140 km until emptying into the Araguari River. The river basin covers an area of approximately 22 thousand km², with altitude between 550 and 1000 m and rainfall of 1,505 mm in 2015. The climate in the region is tropical, with a dry season from May to October and wet season from November to April.

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The river basin is divided into three distinct regions, the upper, middle and lower (see Figure 1). The upper region is located in the northern part of the municipality of Uberlândia and southern part of Uberaba, in an area of sedimentary tableland, containing expansive moist systems in open environments, in the form of natural corridors or fields, at altitudes ranging between 900 and 1000 m, configuring natural water reservoirs that assure perennial drainage. The middle region has areas of moderately dissected topography and vegetation mainly consisting of small patches of riparian forest and fragments of savanna and cerradão (woodland with a well-developed arboreal canopy), covering most of Uberlândia. The lower region has intensely dissected topography in the lower portion of the basin (downstream from Uberlândia), and

the vegetation is characterized by seasonal forest fragments and cerradão, at altitudes varying from 640 to 800 m, where the Uberabinha River empties into the Araguari River (Schneider, 1996).

The population in Uberlândia city is about 650 thousand inhabitants. Uberlândia has three waste treatment stations (WTEs), but 95% of the sewage load is treated by the Uberabinha WTE, which discharges the treated effluent into the river’s lower course (Salla et al., 2014b).

The main uses of the water from the Uberabinha River are human and livestock consumption, irrigation (sugarcane, soybeans and corn), food processing factories and energy generation (Rosolen et al., 2009). The demands from surface and underground water in the basin for human and animal consumption, irrigation and industry in 2015 were 17.07 and 1.61 hm³.month⁻¹, respectively. Of the total surface demand, 14.77 hm³.month⁻¹ is allocated for human consumption, of which 9.59 hm³.month⁻¹ is taken from the Uberabinha River and 5.18 hm³.month⁻¹ from Bom Jardim Stream.

The river basin has two small run-of-the-river (ROR) hydroelectric plants downstream from the discharge point of the Uberabinha WTS, in the

Figure 1. Uberabinha River watershed and monitoring points.

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We selected five monitoring points along the lower course of the Uberabinha River (points 1, 3, 4, 5 and 6) and one point at the outlet of Uberabinha WTS, before confluence with the river (point 2). The segment studied has length of about 19 km, from just upstream from the confluence of the river and the effluent from Uberabinha WTS (identified by point 1 - Figure 1) until the environs of the Martinésia district (identified by point 6 - Figure 1).

2.2. Input data

2.2.1. Water quality

Samples for analysis of water quality were obtained from all points (1 to 6), in the dry season from June to October 2015, to calibrate the coefficients of the physico-chemical reactions, benthic demands, resurgence and diffuse loads belonging to the mathematical equations formulated to represent the self-purification process in this part of the river. The parameters monitored were DO, BOD, NH$_3$, NO$_3$-$N$, P$_{total}$ and the heavy metals Cd, Cr, Cu, Pb and Zn. The techniques for preservation, sampling and laboratory analysis followed the recommendations of the ABNT NBR 9898:1987 (ABNT, 1987).

2.2.2. River flow

The flow of the Uberabinha River was measured only at points 1 and 6, in the dry season from June to October 2015. These points correspond to the start and end of the segment studied. Because of the existence of the Martins ROR power plant near point 5, the flow monitoring at points 1 and 6 was used to calculate the water balance, where:

- in the stretch between point 1 and Martins PP, we estimated the fixed flow as being equal to the sum of the river at point 1 and the effluent from Uberabinha WTS (this was possible because few streams exist along this segment of the river, so their contribution could be considered negligible); and
- in the stretch between Martins PP and point 6, we estimated the fixed flow to be equal to that obtained at point 6 (again possible due to the existence of few streams).

To measure the flow speed, we used an acoustic Doppler current profiler (ADCP) coupled to an echobathymeter calibrated for the river channel cross section. Then a continuity equation was used in the RiverSurveyor computational tool to determine the water speed.

2.2.3. Flow at Uberabinha WTS

The flow at the outlet of Uberabinha WTS, during the dry season from June to October 2015, measured at point 2 (Figure 1), was estimated according to Salla et al. (2013b). The flow was obtained by the flow equation for water consumption distribution multiplied by the coefficient of return, which according to ABNT NBR 9649:1986 (ABNT, 1986) and 14486:2000 (ABNT, 2000) is set at 0.80 when no local empirical data are available.

2.3. Mathematical modeling of the physical and biochemical processes

To study the dispersion and quantify and accompany the evolution over time of the pollutants discharged into the river, we used the advection-diffusion equation, represented in (1).

\[
\frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C = - \mathbf{g} \frac{\partial (\nabla C)}{\partial x_i} \tag{1}
\]

In Equation 1: $\frac{\partial C}{\partial t}$ denotes the variation of concentration in time; $\mathbf{u} \cdot \nabla C$ represents the pollutant transport by advection; $\partial (\nabla C)/\partial x_i$ represents the pollutant transport by turbulent diffusion; $\mathbf{g}$ is the source or sink term (conversions or physical and biochemical processes undergone by the water quality parameter); and $i$ represents the directions x, y and z.

Chapra (1997) suggests the following simplifications of Equation 1 for water quality modeling in lotic environments: permanent flow ($\frac{\partial C}{\partial t}$ equal to zero); unidirectional flow (only longitudinal direction x) on a monthly scale; and negligible turbulent diffusion term. The reasons are: concentration of the pollutant in domestic effluent have little variation throughout the day, which suggests a concentration fixed or permanent over month; most mathematical modeling of self-purification of lotic environment consider a complete and instant mixing of river main and tributary pollution load, or despises the mixing zone (pollutant plume). From the complete mixture, the sidewalls and bottom of the channel act as barrier that prevents the dispersion of the pollutant in y and z directions, enabling the rejection of modeling in y and z directions; in lotic ambient the advective effect prevails over the diffusive effect in behavior of different pollutants, allowing despise the diffusive
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effect. The opposite occurs in lentic environment (storage reservoirs).

With these simplifications, Equation 1 becomes Equation 2.

In Equation 2, \( \frac{\partial C}{\partial t} \) is represented by \( \frac{\partial C}{\partial t} \), and numerical solution by finite difference results in Equation 3.

\[
- \frac{\partial C}{\partial t} = \pm g (t_{n+1} - t_n)
\]

Equation 3: \( C_{n+1} = C_n \pm g (t_{n+1} - t_n) \)

In Equation 2, \( \bar{u} \) is the average velocity. In Equation 3: \( C_{n+1} \) represents the concentration of the parameter in the river at time \( t_{n+1} \) after the start of the modeling and \( C_n \) represents the concentration of the parameter in the river at time \( t_n \) after the start of the modeling.

The term \( g \) for the parameters DO, BOD, \( NH_3 \), \( NO_3 \), \( P_{total} \), Cd, Cr, Cu, Pb and Zn is represented by Equations 4, 5, 6, 7, 8 and 9. We used a spreadsheet program for mathematical solution of these equations, with spatial discretization of 100 m.

\[
\begin{align*}
\text{g(DO)} &= + k_d \cdot 1.024(T-20) \cdot (DO_{sat} - DO) - \\
&= k_d \cdot 1.047(T-20) \cdot BOD - R_{02-ammonia-nitrit} \cdot k_D \cdot 1.080(T-20) \cdot NH_3 \\
&= R_{02-nitrite-nitrit} \cdot k_D \cdot 1.047(T-20) \cdot NO_2^- - S_{an} \cdot h \cdot 2^{-1} \\
\text{g(BOD)} &= - k_d \cdot 1.047(T-20) \cdot BOD + 1 \cdot \text{diffus} \\
&= - k_d \cdot 1.024(T-20) \cdot BOD - R_{02-ammonia-nitrit} \cdot k_D \cdot 1.080(T-20) \cdot NH_3 \\
&= R_{02-nitrite-nitrit} \cdot k_D \cdot 1.047(T-20) \cdot NO_2^- \\
\text{g(NH3)} &= + k_{ox} \cdot 1.047(T-20) \cdot N_{org} \\
&= R_{02-ammonia-nitrit} \cdot k_D \cdot 1.080(T-20) \cdot NH_3 \\
\text{g(NO3 -)} &= + R_{02-nitrite-nitrit} \cdot k_D \cdot 1.047(T-20) \cdot NO_2^- \\
\text{g(Ptotal)} &= - k_p \cdot P_{total} + S_p \\
\text{g(metal)} &= - k_{metal} \cdot C_{metal} + S_{metal}
\end{align*}
\]

Where: \( T \) is water temperature, in °C; \( DO_{sat} \) is dissolved oxygen saturation, in mg.L\(^{-1} \); \( DO \) is dissolved oxygen, in mg.L\(^{-1} \); \( k_s \) is the coefficient of reaeration, in day\(^{-1} \); \( BOD \) is the biochemical oxygen demand, in mg.L\(^{-1} \); \( k_d \) is the coefficient of carbonaceous organic matter decomposition, in day\(^{-1} \); \( L_{diffus} \) is the diffuse BOD load, in g.m\(^{3}.day\(^{-1} \); \( S_j \) is the benthic DO demand, in gO\(_2\).m\(^{2}.day\(^{-1} \); \( h \) is the water depth, in m; \( k_b \) is the coefficient of decay by sedimentation of carbonaceous organic matter, in day\(^{-1} \); \( NH_3 \) is ammoniac nitrogen, in mg.L\(^{-1} \); \( N_{org} \) is organic nitrogen, in mg.L\(^{-1} \); \( NO_3^- \) is nitrite, in mg.L\(^{-1} \); \( NO_2^- \) is nitrate, in mg.L\(^{-1} \); \( k_m \) is the coefficient of conversion of \( NO_3^- \) into \( NO_2^- \), in day\(^{-1} \); \( k_s \) is the coefficient of conversion of \( N_{org} \) into \( NH_3 \), in day\(^{-1} \); \( R_{02-ammonia} \) is the oxygen consumption by oxidation of \( NH_3 \), in mgO\(_2\).mgNH\(_3\) \(^{-1} \); \( R_{02-nitrit} \) is the oxygen consumption for oxidation of \( NO_3^- \), in mgO\(_2\).mgNO\(_2\) \(^{-1} \); \( \phi_{nitrit} \) is the nitrification factor, non-dimensional; \( K_{ox} \) is the coefficient of conversion of \( NH_3 \) into \( NO_3^- \), in day\(^{-1} \); \( P_{total} \) is the total phosphorus, in mg.L\(^{-1} \); \( k_p \) is the coefficient of decay of \( P_{total} \), in day\(^{-1} \); \( S_{an} \) is the release of de \( P_{inorg} \) by the bottom sediment, in g.m\(^{3}.day\(^{-1} \); \( S_{metal} \) is the total metals (cadmium, copper, chromium, lead and zinc), in mg.L\(^{-1} \); \( k_{metal} \) is the coefficient of decay of metals, in day\(^{-1} \); and \( S_{metal} \) is the release of metals by the bottom sediment, in g.m\(^{3}.day\(^{-1} \).

In this study we only analyzed the decay of total phosphorus, whose coefficient of decay was identified by \( k_p \) (day\(^{-1} \)). Therefore, this coefficient encompasses the coefficient of conversion of organic into inorganic phosphorus \( k_w \) (day\(^{-1} \)) and the coefficient of sedimentation of inorganic phosphorus \( k_{ss} \) (day\(^{-1} \)).

The physical and biochemical processes to which the water quality parameters are subject, along with the interaction of the parameters, are presented in Figure 2 for better understanding of the behavior of the water course.

2.4. State variables from the literature

In modeling the longitudinal behavior of the water quality parameters from the state variables available in the literature, we used the averages for \( k_s \), \( k_{an} \), \( k_{ox} \) and \( k_{ss} \) from data for all months. For the coefficient of metal decay \( k_{metal} \), we also employed a fixed value for all months, equal to 0.01 day\(^{-1} \), which is the standard value supplied by the AquaTool computational tool. This tool, used to support water resource planning decisions, recently incorporated the option of modeling various toxic contaminants in its water quality modeling module (GESCAL module), such as heavy metals, emerging organic contaminants and detergents, among others (Salla et al., 2014b; Paredes-Arquiiola et al., 2010).

The coefficients \( k_s \) and \( k_d \) were estimated by equations from the literature, which depend on water level and average runoff and flow speed (see Table 1), so they varied over the months. The hydraulic ratios were obtained from Manning’s equation, assuming:

- The entire 19 km segment of the Uberabinha River has uniform trapezoidal cross section, with channel base width \( b \) ranging from 23 and 27 m and slope of the river banks set at 45°. For this, we performed bathymetric field measurements in October 2015 at the end points of the river segment studied.
- The base material in the channel is composed of basalt and a few small boulders ($\eta_1$ equal to 0.035 m$^{-1/3}$s, according to Chow, 2009) and the banks are covered with mediums-size plant species ($\eta_2$ equal to 0.065 m$^{-1/3}$s, according to Chow, 2009). By the method of Cowan, we used a roughness coefficient of 0.046 m$^{-1/3}$s, assumed to be fixed along the entire river segment. Among the various methods to estimate Manning's roughness coefficient, the method of Cowan is widely used because it considers various factors that influence the flow resistance (Chow, 2009);

- The longitudinal declivity of the channel bottom was obtained from elevation data measured by radar interferometry from the TOPODATA database (INPE, 2015) in the QGIS tool, with horizontal sampling interval of 30 m, subsequently generalized to 100 m. Methods that use GIS are nearer to reality than other less precise methods, such as average declivity in the entire river stretch computed by the starting and ending elevations obtained in the Google Earth tool.

In these simulations, because of the relative scarcity of data in the literature and the fact we only took measurements in the dry season, the resurgence loads $S_p$ and $S_{metal}$ the benthic DO demand $S_d$ and diffuse load $L_{diffus}$ were disregarded.
2.5. Calibration and sensitivity analysis of the coefficients of biochemical, diffusion and resurgence loads

Based on numerical solution by finite difference of the differential Equations 4 to 9, with a discretization step of 100 m, we simulated the profile of each parameter. We then compared these simulated profiles with the values measured at points 3, 4, 5 and 6 and then performed an adjustment so that the simulated values were as close as possible to the empirical values. That adjustment involved calibration of the following state variables: coefficients of biochemical reactions $k_a$, $k_d$, $k_s$, $k_{oa}$, $k_{an}$, $k_{nn}$, $k_P$, $k_{metal}$; diffuse load $L_{diff}$; and resurgence loads $S_{metal}$ and $S_p$. The benthic demand $S_d$ was measured in the laboratory by the method described next.

The process involved trial and error, starting by attributing average values from the literature as the initial calibration step, attributing fixed values in a stretch with the same longitudinal declivity. We identified five stretches along the portion of the Uberabinha River studied, with the separation criterion being proximity of longitudinal declivity. Figure 3 identifies the river stretches and monitoring points.

The sharpest changes in longitudinal declivity occur in stretches 1 and 3. In stretch 1 the declivity is 3.67%, while in stretch 3 it is 6.09% due to the existence of the Martins power plant. Stretches 2 and 5 have the smallest longitudinal declivities, equal to 0.47% and 0.29%, respectively. Stretch 4 has intermediate declivity of 0.79%.

The number of state variables calibrated in this study is large. Sensitivity analysis of the coefficients, the resurgence loads and benthic DO demand allow assessing the relative importance of each state variable in the behavior of the various water quality parameters. The interactions of these parameters are illustrated in Figure 2. In each river stretch we quantified the average percentage variation of the diverse parameters based on oscillation by +10% and -10% of all the coefficients, resurgence loads and benthic BO demand. In each simulation, while one variable fluctuated +10% or -10%, the others were held steady at the original calibrated value.

2.6. Quantification of the benthic DO demand

Quantification of the benthic DO demand ($S_d$) was necessary to reduce the number of coefficients to be calibrated, since this term is highly sensitive in the behavior of the parameters DO and BOD (Salla et al., 2014b). The benthic DO demand $S_d$ (in gO$_2$.m$^{-2}$.day$^{-1}$) was quantified at five points (points 1, 3, 4, 5 and 6), by collecting a water sample from just above the river bed. In the laboratory, the samples were stored in the bottles of an OxiTop® IS 6 system under controlled temperature conditions (20 °C). With the daily values of DO consumption inside each bottle and considering the internal cross section as the surface area, it was possible to estimate the benthic DO demand by applying Equation 10.

$$S_d = \left[ (\text{BOD}_0 - \text{BOD}_1) \times 10^{-3} \times A^{-1} \right]$$

Where: $S_d$ is the benthic DO demand (gO$_2$.m$^{-2}$.day$^{-1}$); $\text{BOD}_0$ is the remaining BOD at time zero (mg.L$^{-1}$); $\text{BOD}_1$ is the remaining BOD at the end of the first day (mg.L$^{-1}$); and $A$ is the internal bottle cross section area (m$^2$), set at 0.00442 m$^2$.

Figure 3. Identification of stretches and monitoring points.
The samples were collected at the interface between the liquid portion and the river bottom using a bottom collector, which allowed obtaining a sample from the water-solid interface without loss or contamination of the sample while bringing it to the surface.

2.7. Comparison of the water quality model with calibrated state variables and those obtained in the literature

Because of the scarcity of historical water quality data in Brazil, many studies of self-purification in lotic environments estimate the state variables based on fixed values and empirical equations from the literature, related to the morphological and hydraulic characteristics of the channel, previous river water quality and quality of the effluent released. However, according to Salla et al. (2014a), mathematical modeling satisfactorily reflects the self-purification process when these variables are calibrated against water quality data from monitoring the stretch of the river of interest.

To demonstrate the importance of calibration in the mathematical modeling of self-purification, we compared the longitudinal behavior of the water quality parameters studied here after calibration/measurement of the state variables with the longitudinal behavior of the same parameters obtained from state variables in the literature

3. Results and Discussion

3.1. Input data

The flow and water quality data obtained by monitoring are reported in Table 2. The water quality data from points 1 and 2 and the flow from points 1 and 6 were used as input for the simulations, while the water quality data from points 3 to 6 were employed in calibrating the state variables.

The focus of this study is evaluation of the importance of calibrating a water quality model in a lotic environment in the dry season. According to data from the National Meteorology Institute (INMET, 2016), collected at the Uberlândia Automatic Surface Weather Observation Station, the cumulative monthly rainfall figures in 2015 were 25.4 mm in June, 8.2 mm in July, zero in August, 46.2 mm in September and 86.0 mm in October.

The field work was performed in days without rain from June to September 2015. In October 2015, the weather station on the collection day indicated the occurrence of precipitation, without

Table 2. Flow and water quality data at six monitoring points.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Point 1</th>
<th>Point 2</th>
<th>Point 3</th>
<th>Point 4</th>
<th>Point 5</th>
<th>Point 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Flow (m³.s⁻¹)</td>
<td>4.2±0.8</td>
<td>2.0±0.2</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>4.8±0.9</td>
</tr>
<tr>
<td>DO (mg.L⁻¹)</td>
<td>3.3±6.2</td>
<td>0.0±0.0</td>
<td>1.5±5.2</td>
<td>1.2±3.4</td>
<td>0.7±2.7</td>
<td>2.4±7.1</td>
</tr>
<tr>
<td>BOD₅ (mg.L⁻¹)</td>
<td>3.0±34.0</td>
<td>98.0±174.0</td>
<td>25.0±95.0</td>
<td>9.0±125.0</td>
<td>7.0±138.0</td>
<td>6.0±137.0</td>
</tr>
<tr>
<td>N₅ (mg.L⁻¹)</td>
<td>0.20±0.30</td>
<td>19.0±20.0</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>NH₃ (mg.L⁻¹)</td>
<td>0.02±0.40</td>
<td>0.08±0.20</td>
<td>0.03±0.14</td>
<td>0.03±0.14</td>
<td>0.03±0.14</td>
<td>0.03±0.14</td>
</tr>
<tr>
<td>NO₂⁻ (mg.L⁻¹)</td>
<td>0.44±1.26</td>
<td>0.02±0.75</td>
<td>0.41±1.19</td>
<td>0.06±0.35</td>
<td>0.03±0.23</td>
<td>0.75±2.74</td>
</tr>
<tr>
<td>NO₃⁻ (mg.L⁻¹)</td>
<td>0.00±0.00</td>
<td>0.00±0.00</td>
<td>0.00±0.00</td>
<td>0.00±0.00</td>
<td>0.00±0.00</td>
<td>0.00±0.00</td>
</tr>
<tr>
<td>P (mg.L⁻¹)</td>
<td>0.26±2.39</td>
<td>0.36±4.63</td>
<td>1.10±4.59</td>
<td>1.40±4.34</td>
<td>0.98±4.59</td>
<td>1.23±4.63</td>
</tr>
<tr>
<td>Cd (mg.L⁻¹)</td>
<td>0.12±0.20</td>
<td>0.04±0.43</td>
<td>0.08±0.20</td>
<td>0.06±0.24</td>
<td>0.01±0.21</td>
<td>0.01±0.06</td>
</tr>
<tr>
<td>Cr (mg.L⁻¹)</td>
<td>0.00±1.76</td>
<td>0.00±1.78</td>
<td>0.00±1.72</td>
<td>0.00±2.20</td>
<td>0.00±2.20</td>
<td>0.00±2.20</td>
</tr>
<tr>
<td>Cu (mg.L⁻¹)</td>
<td>0.07±0.08</td>
<td>0.08±0.08</td>
<td>0.08±0.05</td>
<td>0.04±0.02</td>
<td>0.05±0.02</td>
<td>0.02±0.02</td>
</tr>
<tr>
<td>Pb (mg.L⁻¹)</td>
<td>0.00±6.55</td>
<td>0.00±2.92</td>
<td>0.00±5.20</td>
<td>0.00±2.60</td>
<td>0.00±1.59</td>
<td>0.00±1.59</td>
</tr>
<tr>
<td>Zn (mg.L⁻¹)</td>
<td>0.21±5.01</td>
<td>0.58±2.85</td>
<td>0.63±4.20</td>
<td>0.39±2.10</td>
<td>0.33±1.20</td>
<td>0.29±1.60</td>
</tr>
</tbody>
</table>

Minimum-maximum; Mean ± standard deviation.
Identification of the collection points affected. This caused the need to consider diffuse pollutant loads for that month in the mathematical modeling of self-purification.

With respect to the water quality results for class 2 watercourses, in which the Uberabinha River is classified, discrepancies were found with the limits defined in CONAMA (National Environmental Council) Resolutions 357/2005 (BRASIL, 2005) and 430/11 (BRASIL, 2011) and COPAM (State Environmental Policy Council) Normative Deliberation 01/2008 (BRASIL, 2008), specifically for the parameters DO, BOD, P\text{total} and the metals cadmium, chromium, lead and zinc. Regarding copper, it was not possible to compare the observed values with the regulatory limits since the latter limits refer to the maximum concentration of dissolved copper and the analyses in this study measured total copper. To clarify, the limits defined in the resolutions for class 2 rivers are: minimum of 5.0 mg.L\(^{-1}\) for DO, maximum of 5.0 mg.L\(^{-1}\) for BOD, maximum of 0.1 mg.L\(^{-1}\) for P\text{total}, maximum of 0.001 mg.L\(^{-1}\) for cadmium, maximum of 0.05 mg.L\(^{-1}\) for chromium, maximum of 0.01 mg.L\(^{-1}\) for lead and maximum of 0.18 mg.L\(^{-1}\) for zinc.

At point 1, located upstream from the Uberabinha WTS, the surface water quality was degraded by clandestine discharge of sewage into local streams flowing into the Uberabinha River. The same finding was reported by Rosolen et al. (2009).

The assessment of metals is necessary because at high concentrations they can cause negative effects on the environment and living things.

Cadmium, used to galvanize steel and as an anticorrosion agent, is released in watercourses by disintegration of rocks, erosion or fluvial transport. The frequent ingestion of this metal in seafood such as oysters, scallops, mollusks and crustaceans can alter the metabolism of calcium in the liver and kidneys of mammals, weakening the bones and promoting osteoporosis (Harte et al., 1991).

Lead, found in batteries, rain gutters, paints, ceramics, roof tiles and X-ray radiation protectors, has low solubility in water and is mainly transported by air, so this is the main route by which plants absorb this toxic substance (Parsons & Dixon, 2014). Spices, cereals and canned foods generally have the highest levels of lead (Intawongse & Dean, 2006). One-off exposure to high concentrations can cause gastrointestinal disorders, renal and hepatic lesions, hypertension and neurological effects. Prolonged exposure to lower levels can cause anemia, headaches, irritability, lethargy, convulsions, muscular weakness, ataxia, tremors, paralysis and cancer (Harte et al., 1991).

Copper, from mining, metal casting, manufacture of electrical wiring, metal alloys and runoff from pesticides, is usually found in natural aquatic systems in small concentrations and generally in cupric form (Cu\(^{2+}\)), linked to carbonates, cyanides, amino acids and other chemical substances. However, in sediments it is deposited in the form of hydroxides, phosphates and sulfates (Shotyk & Le Roux, 2005). Copper can penetrate biological tissues by different pathways, such as oral, inhalation or cutaneous absorption. In humans, the ingestion of salts of this metal can cause gastrointestinal and renal pathologies, with varied symptoms, ranging from deep abdominal pains, vomiting, hepatic necrosis, hypertension, tachycardia, convulsion and even death.

Chromium is released into the environment by producers of dyes, lacquers and paints, leather tanneries and steel alloy makers. It is not found in nature in its elementary form, but rather is combined with other minerals. Chromium is not quickly absorbed, but absorption of Cr (VI) is faster (Sá Ferreira, 2001). Its toxicity is usually manifested by renal problems in humans (Lamson & Plaza, 2002), mutagenic effects (Qian, 2004) and germination inhibition of seeds, stunted growth of plants (Yogeetha et al., 2004) and peroxidation of lipids associated with the formation of reactive oxygen species in plants.

Zinc is an essential metal found in some foods, soils and volcanic rocks, as well as in drinking water in the form of salts or organic complexes. The daily zinc requirement for human adults is 15 to 20 mg, and the main sources of the element are meat, fish, poultry, dairy products, cereals, grains, vegetables and fruits. This metal also has protective action against the toxic effects of cadmium and lead, but the exaggerated ingestion of foods containing zinc can cause colic, nausea, vomiting, diarrhea and fever.

3.2. Estimated and calibrated state variables

Table 3 reports the calibrated and estimated values from the literature, for each stretch of the river, of the coefficients of the biochemical reactions, diffuse load and resurgence, besides the benthic DO demand measured in the laboratory, considered in the water quality modeling of the Uberabinha River.

According to Table 3, the values estimated by the literature for the coefficients \(k\), \(k_p\), \(K_c\), \(K_{oa}\), \(K_{an}\) and \(K_p\),
Table 3. Calibrated and estimated values of the coefficients of biochemical reactions, diffuse load and demand/resurgence.

<table>
<thead>
<tr>
<th></th>
<th>Estimated (literature)</th>
<th>Calibrated</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Stretch 1</td>
<td>Stretch 2</td>
</tr>
<tr>
<td>$ka$ (day$^{-1}$)</td>
<td>96.7-140.8</td>
<td>5.4-31.2</td>
</tr>
<tr>
<td></td>
<td>119.5±17.3</td>
<td>17.4±9.4</td>
</tr>
<tr>
<td>$kd$ (day$^{-1}$)</td>
<td>0.87-1.18</td>
<td>0.53-0.91</td>
</tr>
<tr>
<td></td>
<td>0.97±0.11</td>
<td>0.68±0.11</td>
</tr>
<tr>
<td>$ks$ (day$^{-1}$)</td>
<td>0.02</td>
<td></td>
</tr>
<tr>
<td>$koa$ (day$^{-1}$)</td>
<td>0.20</td>
<td></td>
</tr>
<tr>
<td>$kna$ (day$^{-1}$)</td>
<td>0.20</td>
<td></td>
</tr>
<tr>
<td>$kP$ (day$^{-1}$)</td>
<td>0.29</td>
<td></td>
</tr>
<tr>
<td>$SP$ (g.m$^{-3}$.day$^{-1}$)</td>
<td></td>
<td>2.19-84.50</td>
</tr>
<tr>
<td>$Sd$ (g.m$^{-3}$.day$^{-1}$)</td>
<td>-</td>
<td>32.99±28.97</td>
</tr>
<tr>
<td>$Ldif$ (g.m$^{-3}$.day$^{-1}$)</td>
<td></td>
<td>6.80-84.96</td>
</tr>
<tr>
<td>$kCd$ (day$^{-1}$)</td>
<td>0.01</td>
<td></td>
</tr>
<tr>
<td>$Sd$ (g.m$^{-3}$.day$^{-1}$)</td>
<td>-</td>
<td>8.73±28.50</td>
</tr>
<tr>
<td>$Ldif$ (g.m$^{-3}$.day$^{-1}$)</td>
<td></td>
<td>28.30±29.10</td>
</tr>
<tr>
<td>$kCu$ (day$^{-1}$)</td>
<td>0.20</td>
<td>-</td>
</tr>
<tr>
<td>$Scu$ (g.m$^{-3}$.day$^{-1}$)</td>
<td>-</td>
<td>0.01</td>
</tr>
<tr>
<td>$kCr$ (day$^{-1}$)</td>
<td>0.01</td>
<td>-</td>
</tr>
<tr>
<td>$Scr$ (g.m$^{-3}$.day$^{-1}$)</td>
<td>-</td>
<td>0.00±0.18</td>
</tr>
<tr>
<td>$kB$ (day$^{-1}$)</td>
<td>0.01</td>
<td>-</td>
</tr>
<tr>
<td>$SPb$ (g.m$^{-3}$.day$^{-1}$)</td>
<td>-</td>
<td>0.22±0.03</td>
</tr>
<tr>
<td>$Zn$ (day$^{-1}$)</td>
<td>0.01</td>
<td>-</td>
</tr>
<tr>
<td>$SZn$ (g.m$^{-3}$.day$^{-1}$)</td>
<td>-</td>
<td>0.00-30.0</td>
</tr>
<tr>
<td>$Sd$ (g.m$^{-3}$.day$^{-1}$)</td>
<td>-</td>
<td>6.08±12.25</td>
</tr>
<tr>
<td>$Ldif$ (g.m$^{-3}$.day$^{-1}$)</td>
<td></td>
<td>4.59±5.46</td>
</tr>
</tbody>
</table>

Minimum-maximum; Mean ± standard deviation.
follow the recommendations of Chapra (2003) and Von Sperling (2007). Similar values were used by Ferreira (2014) and Salla et al. (2015). With the exception of $k_d$ and $k_s$, which were obtained from the hydraulic relations, all the other estimated coefficients were held constant in the period from June to October 2015.

The values estimated for $k_{aa}$, $k_{ac}$, $k_{ab}$, $k_{ap}$ and $k_{ao}$ were assumed to be fixed and equal to 0.01 day$^{-1}$, according to Paredes-Arquiola et al. (2010). The estimated reburial loads of metals ($S_{Ca}$, $S_{Cu}$, $S_{Pb}$ and $S_{zn}$) and phosphorus ($S_p$) and benthic DO demand ($S_d$) were considered nil based on the lack of data in the literature. The estimated diffuse load ($L_{diffu}$) was disregarded because the study was carried out in the dry season.

The comparison between the calibrated state variables with those found in the literature brings some important information about how adequate the water quality modeling is in lotic environments based on state variables obtained from the literature. The discrepancies found between the estimated and calibrated values are associated with the low capacity for dilution of pollutants in the dry season and the high pollutant load discharged by the Uberabinha WTS.

Regarding the coefficient of natural reaeration ($k_a$), the equations in the literature overestimate its value. Using stretch 2 as an example, $k_a$ from the literature is on average four times the calibrated value.

Likewise, the equations in the literature overestimate the coefficient of decomposition of carbonaceous organic matter ($k_d$). Again employing stretch 2 as an example, $k_d$ from the literature is on average about 1.6 times the calibrated value. This fact shows the need to calibrate $k_d$ and $k_d$ based on water quality data monitored along the entire course of the river studied.

In the calibration process, the coefficient of decay by sedimentation of carbonaceous organic matter in the river ($k$) was higher than the value of 0.02 day$^{-1}$ recommended by the literature (average of 0.40 day$^{-1}$ in June 2015, 1.42 day$^{-1}$ in July 2015, 1.40 day$^{-1}$ in August 2015, 1.40 day$^{-1}$ in September 2015 and 0.03 day$^{-1}$ in October 2015, as shown in Table 3).

For the ammonification process, the estimated and calibrated values in the dry season of the coefficient of conversion of organic nitrogen into ammonia ($K_{aa}$) were not close (0.20 day$^{-1}$ estimated and 0.001 day$^{-1}$ calibrated). Regarding nitrification, fixed values of $k_{nn}$ equal to 0.20 day$^{-1}$ and $k_{nn}$ equal to 0.75 day$^{-1}$ were used in the simulations based on coefficients from the literature. The calibrated coefficients $k_{nn}$ were between 0.40 and 0.90 day$^{-1}$ and $k_{nn}$ were between 0.00 and 0.05 day$^{-1}$ (Table 3). In this study, the low calibrated values of $k_{nn}$ are directly related to the low concentrations of nitrate in the water.

Regarding the total phosphorus parameter, the average value of the calibrated coefficient of decay ($k_d$) was lower than that estimated in the literature in all the stretches (0.29 day$^{-1}$ estimated and 0.13 day$^{-1}$ calibrated). The calibrated values of release of $P_{inorg}$ by the bottom sediment ($S_p$) were on average 32.99 g.m$^{-3}$.day$^{-1}$ in stretch 1, 2.03 g.m$^{-3}$.day$^{-1}$ in stretch 2, 0.70 g.m$^{-3}$.day$^{-1}$ in stretch 3, 0.38 g.m$^{-3}$.day$^{-1}$ in stretch 4 and 0.26 g.m$^{-3}$.day$^{-1}$ in stretch 5. The tendency for these values do decline in stretches 3, 4 and 5 is related to the capacity for sedimentation of inorganic phosphorus in stretch 2, specifically near the reservoir of the Martins PP.

The calibrated diffuse BOD load ($L_{diffu}$) was only considered in October de 2015, the rainiest month in the period studied. This metric represents the diffuse surface entry of carbonaceous organic matter along the river course, equal to 207.3, 69.1, 30.4, 16.1 and 11.0 g.m$^{-3}$.day$^{-1}$, respectively, in stretches 1, 2, 3, 4 and 5. The high values in stretches 1 and 2 are directly related to the land use and occupation in the region: urbanization in stretch 1 and use of fertilizers and pesticides on various crops grown in the experimental farm of Uberlândia Federal University in stretch 2.

The benthic DO demand ($S_d$) values measured in the laboratory ranged from 2.0 to 18.5 g.m$^{-2}$.day$^{-1}$ in stretch 1, 2.0 to 2.8 g.m$^{-2}$.day$^{-1}$ in stretch 2 and 2.4 to 2.8 g.m$^{-2}$.day$^{-1}$ in stretches 3, 4 and 5. The literature in this respect reports values close to these in watercourses with similar morphological and water quality characteristics. For example, Thomann and Mueller (1987) reported ranges from 1.0 to 2.0 g.m$^{-2}$.day$^{-1}$ and Penteado (2009) observed 0.5 to 7.0 g.m$^{-2}$.day$^{-1}$. For comparison, in watercourses with good quality because they do not receive considerable sewage loads, the reported benthic DO demand values are lower, on the order of 0.10 to 1.4 g.m$^{-2}$.day$^{-1}$ (Edeberg & Hofsten, 1973; Salla et al., 2014b).

Salla et al. (2014b), in analyzing the water quality of the same river, in a segment approximately 48 km long of the upper, middle and lower course, in the period from October 2006 to September 2011, obtained the following calibrated values of coefficients: $k_a$ (0.08 day$^{-1}$), $k_d$ (0.05-0.06 day$^{-1}$),

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for lead; and $K_{-1}$ metal values were between L, we believe the discrepancies of the $k_{-1}$ inorg. day$^{-1}$ Superville et al., 2015 relation exists between higher values of chromium, which has low adsorption capacity. For stretches 1 to 5, the average coefficient of decomposition (k) was opposite for release of P$_{b}$, with the exception of the nitrate parameter, where decreases the accumulation of bottom sediment), prevails over the effects of longitudinal declivity. According to Table 3, the coefficients from the literature and calibrated ones used for ammonification and nitrification (K$_{an}$, k$_d$ and k$_{an}$) were not influenced significantly by the change in longitudinal declivity in the five stretches. The same situation was observed for decay of total phosphorus (k$_d$), decay of cadmium and chromium (K$_{Cd}$ and K$_{Cr}$) and release of cadmium and chromium by the bottom sediment (S$_{Cd}$ and S$_{Cr}$). The situation was opposite for release of P$_{bb}$ by the bottom sediment (S$_{Pb}$), for decay of copper, lead and zinc (k$_{Cu}$, k$_{Pb}$ and k$_{Zn}$) and release of copper, lead and zinc by the bottom sediment (S$_{Cu}$, S$_{Pb}$ and S$_{Zn}$).

3.3. Simulation profiles

As shown by Figure 4, the time series simulated by the calibration process and the data measured at points 3, 4, 5 and 6 presented good fits according to the Nash-Sutcliffe coefficient for the parameters DO (coefficient of 0.91 to 0.99), BOD (0.96 to 0.99), ammonia (0.83 to 0.98), total phosphorus (0.90 to 1.00), cadmium (0.83 to 0.91), lead (0.84 to 0.99), copper (0.70 to 0.99), zinc (0.81 to 0.93) and chromium (0.97 to 0.98). The exception was the nitrate parameter, where the time series were above the values measured in all months studied, but with the same monthly variation tendency.

Figure 5 compares the longitudinal profiles of the simulated parameters based on the calibrated state variables with the simulated profiles based on the state variables obtained from the literature, along with the water quality data monitored at points 3 to 6. In this figure, comparison of the longitudinal profiles of the simulations based on the calibrated values with the measured water quality data shows good adjustments between the simulated and measured data for all the parameters analyzed. The comparison of the water quality model using
calibrated data with the model using variables obtained from the literature reveals inconsistencies for the parameters DO, $P_{\text{total}}$, ammonia, nitrate and all the heavy metals.

The sharp discrepancies of the longitudinal profiles for DO are related to the coefficient of natural reaeration ($k_a$), which is overestimated by the equations from the literature, as mentioned previously.

With respect to BOD, although the equations from the literature overestimate the value of the coefficient of decomposition of carbonaceous
Figure 5. Longitudinal profiles of the simulated water quality parameters and water quality data monitored at points 3 to 6.

The explanation for the discrepancy in the profiles of the total phosphorus parameter (see Figure 5) is the omission of the bottom resurgence of inorganic phosphorus in the simulations based on values from the literature. Even though the coefficient of decay of total phosphorus (which encompasses the conversion of organic into inorganic phosphorus

organic matter in the river \( (k) \) by an average of approximately 1.6 times the calibrated value, Figure 5 shows a certain proximity between the longitudinal profiles for this parameter. The only discrepancy in the BOD profile was in October 2015, due to the diffuse load \( (L_{\text{diff}}) \) considered in stretches 1 and 2.
and sedimentation of inorganic phosphorus) in the calibration was about double the value from the literature, the bottom resurgence of inorganic phosphorus prevailed in the model for total phosphorus, mainly in stretch 1.

The accentuated discrepancies between the profiles of the ammonia and nitrate parameters (see Figure 5) are related to the high values of the coefficients from the literature involving ammonification \( (K_a, k_a) \) equal to 0.20 day\(^{-1}\). To support this affirmation, Table 3 shows that the sum of the coefficients involved in nitrification \( (K_{an}, K_{an}) \) from literature data and the calibration are close in all five stretches.

In the simulations of the heavy metals, in the calibration process the resurgence load tended to prevail over the decay of these metals due to the high longitudinal declivity of stretch 1. Even though the data were collected in the dry season, the high longitudinal declivity in this stretch caused resuspension of the metal absorbed in inorganic matter. In contrast, in stretches 2, 3, 4 and 5 the decay of copper, zinc, lead and cadmium prevailed over the resurgence from the bottom.

With respect to the metals chromium and lead in June and July 2015, the low concentrations found in the river allowed good proximity between the simulated profiles, irrespective of the values of the coefficient of decay of metals and the resurgence load considered. In turn, in August, September and October 2015, the resurgence loads tended to prevail over the decay of chromium in stretches 2, 3, 4 and 5.

The discrepancies in the profiles of the metals can be explained by the disregard in the simulations from literature values of the bottom resurgence of metals and the consideration of a single coefficient of decay for all the metals, equal to 0.01 day\(^{-1}\).

The discrepancies discussed above add further evidence of the need for constant monitoring of water quality for adequate calibration of mathematical models of self-purification.

### 3.4. Sensitivity analysis

We also analyzed the sensitivity of the behavior of the water quality parameters based on fluctuations of +10% and -10% in the calibrated values of the state variables \( k_a, k_{an}, k_s, k_d, k_{Ca}^r, k_{Cd}^r, k_{Zn}^r, k_{Pb}^r, k_{Cu}^r, k_{S}^r, k_{P}^r, k_{N}^r, k_{S}^r, k_{Pb}^r, k_{Cu}^r, k_{Zn}^r, k_{S}^r, k_{P}^r, k_{N}^r, k_{S}^r \), and \( S_{am} \) and \( S_{an} \).

The highest average percentages of variation of the DO occurred with the oscillations of \( k_a, k_d \), and \( S_{p} \) which were, respectively, 5.5%, 2.0% and 1.8% in June, 2.7%, 1.9% and 1.0% in July, 5.3%, 2.8% and 4.5% in August, 7.8%, 3.4% and 2.6% in September, and 29.7%, 16.7% and 3.7% in October. These results show that in October, when rainfall of 32 mm occurred on the collection day, the increase in flow of the river considerably increased the sensitivity in the fluctuation of the coefficients \( k_a, k_d \). In the dry season in general, oscillations of \( k_a, k_p, k_{an}, k_{an} \) and \( S_{p} \) caused little variation in the concentration of DO.

With respect to the BOD parameter, the highest average variation (21.4%) occurred with the fluctuation of \( k_b \) in June. For the other months, no variation was observed in function of the disregard of \( k_b \). The fluctuation of \( k_b \) showed a very small variation of BOD, with maximum values of 1.7% in June, 0.8% in July, 0.9% in August, 1.1% in September and 1.0% in October.

Although to a lesser extent, the highest variation percentages of the ammonia and nitrate parameters occurred in function of the oscillations of the coefficients \( k_{an} \) and \( k_{an} \), respectively. The fluctuations of \( k_{an} \) and \( k_{an} \) were 2.3% and 5.5% in June, 1.8% and 6.2% in July, 1.8% and 4.4% in August, 1.9% and 3.0% in September, and 1.5% and 3.6% in October.

Only the fluctuation of \( S_{p} \) affected the variation of total phosphorus, with figures of 8.6% variation in June, 3.5% in July, 4.7% in August, 0.8% in September, and 5.5% in October. Similar results for the parameters DO, BOD, total phosphorus, ammonia and nitrate were obtained by Salla et al. (2014a, b).

The oscillations of \( k_{metal} \) and \( S_{metal} \) did not follow the same pattern as the variations of cadmium, chromium, copper, lead and zinc. Only the fluctuation of \( k_{metal} \) caused variations of cadmium, equal to 22.7% in June, 4.9% in July, 3.5% in August, 12.9% in September, and 5.9% in October. In counterpart, only the fluctuation of \( S_{metal} \) varied for chromium, equal to 2.5% in June, 2.1% in July, 22.2% in September, and 15.5% in October.

For copper, the oscillations of \( k_{metal} \) (in all months) and \( S_{metal} \) (only in August and September) caused variations in the concentration of this metal. The fluctuation of \( k_{metal} \) caused variations of copper of 16.5% in June, 11.7% in July, 18.5% in August, 8.9% in September, and 10.1% in October. Also, the fluctuation of \( S_{metal} \) varied for copper by 8.8% in August and 9.1% in September.

Similar to the results for copper, the oscillations of \( k_{metal} \) and \( S_{metal} \) caused impacts on the concentration of lead in dispersed form. The fluctuation of \( k_{metal} \) varied for this metal by 11.4% in June and 13.1% in October.
in July, while the fluctuation of $S_{metal}$ varied only in October, by 19.5%.

For zinc, the oscillations of $k_{metal}$ (in all months) and $S_{metal}$ (in June, August and September) caused variations. The fluctuation of $k_{metal}$ caused variations of 20.3% in June, 12.0% in July, 2.3% in August, 11.2% in September, and 11.8% in October. Finally, the fluctuation of $S_{metal}$ varied for zinc by 5.9% in June, 3.7% in August and 10.8% in September. The months without sensitivity can be explained by the low concentration of this metal.

4. Conclusions

The mathematical modeling of self-purification of the Uberabinha River, in the municipality of Uberlândia, Minas Gerais, impacted by the release of effluents from the Uberabinha waste treatment station, was successful. Although not the main focus of this study, it is important to stress that in the period studied the values of the parameters DO, BOD, $P_{total}$ and the levels of the heavy metals cadmium, chromium, lead and zinc were above the limits defined by CONAMA Resolutions 357/2005 (BRASIL, 2005) and 430/11 (BRASIL, 2011) and COPAM Normative Deliberation 01/2008 (BRASIL, 2008) for class 2 watercourses.

The calibration process allowed good fits between the simulated and measured data for all the parameters analyzed. In counterpart, the comparison of the water quality model based on the calibrated state variables and those obtained in the literature revealed inconsistencies for the parameters DO, $P_{total}$, ammonia, nitrate and all the heavy metals.

The sensitivity analysis showed that the most important state variables were $k_1$, $k_2$ and $S_j$ for DO, $k$ for BOD, $S_j$ for total phosphorus, $k_{nn}$ for nitrate, $k_{metal}$ and $S_{metal}$ for heavy metals. No state variables were important to variation of ammonia.

In general, the results reinforce the importance of calibration in mathematical modeling of self-purification, principally in lotic environments that receive high pollution loads in the dry season.

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