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Modelling of trace metal transfer in a large river under different hydrological conditions (the Garonne River in southwest France)

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Abstract

The modelling of trace metals (TM) in rivers is highly dependent on hydrodynamics, the transport of suspended particulate matter (SPM) and the partition between dissolved and particulate phases. A mechanistic, dynamic and distributed model is proposed that describes the fate of trace metals in rivers with respect to hydrodynamics, river morphology, erosion-sedimentation processes and sorption-desorption processes in order to identify the most meaningful parameters and processes involved at the reach scale of a large river. The hydraulic model is based on the 1-D Saint Venant equation integrating real transects to incorporate the river’s morphology. The transport model of dissolved species and suspended sediments is based on advection–dispersion equations and is coupled to the one-dimensional transport with inflow and storage (OTIS) model, which takes transient storage zones into account. The erosion and sedimentation model uses Partheniades equations. Finally, the transfer of trace metals is simulated using two parameters, namely the partition coefficient (Kd) and the concentration of TM in the eroded material.

The model was tested on the middle course of the Garonne River, southwest France, over an 80 km section under two contrasting hydrological conditions (80 m³ s⁻¹ and 800 m³ s⁻¹) based on measurements (hydrology, suspended sediments, particulate and dissolved metals fractions) taken at 13 sampling stations and tributaries. The hydrodynamic model was calibrated with discharge data for the hydraulic model, tracer experiments for the dissolved transport model and SPM data for the erosion-sedimentation model. The TM model was tested on two trace metals: arsenic and lead. Arsenic was chosen for its large dissolved fraction, while lead was chosen for its very important particulate fraction, thus providing contrasting elements. The modelling of TM requires all four processes to be simulated simultaneously. The presented model requires the calibration of ten parameters divided in four submodels during two hydrological conditions (low and high flow). All parameters could be explained by the physical properties of the case study, suggesting that the model could be applied to other case studies. The strategy of using different datasets under different hydrological conditions highlights: (a) the importance of transient storage in the study case, (b) a detailed description of the erosion and sedimentation processes of SPM, and (c) the importance of TM eroded from the sediment as a secondary delayed source for surface water.

1. Introduction

Trace metals (TM) are a class of pollutant that can put ecosystems under pressure. This is particularly the case in fluvial systems since these are the major transport paths from land to sea.

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The transfer of TM is primarily affected by the river hydrology (Shafer et al., 1997; Lanceleur et al., 2011). Other processes can also significantly affect the path and fate of TM, such as physicochemistry (Christensen and Christensen, 2000; Tipping et al., 2003; Cao et al., 2006), organic matter (Tipping, 1994; Ladouche et al., 2001; Garnier et al., 2006) and erosion-sedimentation (Ji et al., 2002; Lindenschmidt et al., 2005). However, few studies focus on the role of hydromorphology in the fate of TM. Hydromorphology primarily includes the influence of erosion and sedimentation zones. It also
includes storage zones, such as dead arms or the hyporheic zone, and complex stream geometry, such as riffle-pool sequences. For example, Bonvallet et al. (2001) demonstrated the importance of hydromorphology in the retention of phosphorus, especially in the hyporheic zone. This retention was closely linked to the river’s hydrology and morphology.

In river waters, TM travel either in dissolved or particulate form based on filtration criteria (Dai and Martin, 1995; Garnier et al., 2006). Therefore, their fate is highly dependent on the fate of suspended particulate matter (SPM) which, in turn, is highly linked to hydrology and morphology through erosion and sedimentation processes.

As TM dynamics are complex, modelling trace metals in river systems requires a proper description of several processes, for each of which different approaches can be used. To simulate the river’s flow, hydrodynamic models are usually based on Saint Venant equations, either in their simplified form, as used for example in the SWAT model (Neitsch et al., 2005) or in their complete form, as used for example by WASP5 (Ambrose et al., 1993), CE-QUAL2 (Brunner, 1995), HEC-RAS (Cole and Wells, 2008) or the Mohid River Network (Trancoso et al., 2009). The transport of suspended elements can then be calculated by mass balance over each river segment (the SWAT approach) or through the use of the advection and dispersion equations (the WASP5, CE-QUAL2, HEC-RAS, Mohid River Network models). However, the advection and dispersion equations are not always sufficient to describe fully the transport of a suspended element in natural rivers due to their heterogeneity (Haggerty et al., 2002; Gooseff et al., 2005; Zarnetske et al., 2007). Therefore, various transport models have been developed to quantify this heterogeneity by including transient storage, such as the one-dimensional transport with inflow and storage (OTIS) model (Bencala and Walters, 1983; Runkel and Broshears, 1991) or the variable residence time model (VART) (Deng and Jung, 2009). The erosion and sedimentation processes, controlling the fate of SPM, are also well documented and modelled. The two main approaches include “transport capacity”, as defined by Bagnold (1977) and used in the SWAT model, and the Partheniades equations (Partheniades, 1965), relying on the computed shear stress on the riverbed. Finally, TM can migrate from the dissolved phase to the particulate phase and vice versa. If a modelling choice can be to neglect this dynamic, the most common description of this dependency is through the coefficient of partition Kd, a fixed separation coefficient. The most important parameters and processes involved at the reach scale were then identified. The middle course of the Garonne River (southwest France) was used as the case study.

2. Materials and methods
2.1. Model development

The trace metal model includes four different submodels. They are included in order of dependency. Therefore, each submodel depends on those described before and is assumed independent of those that follow. The order of these submodels is:

1. Hydrodynamic submodel.
2. Solute transport submodel.

For example, the solute transport needs inputs from the hydrodynamic model, but is independent of the erosion-sedimentation and TM transport models. The processes are schematised in Fig. 1. Therefore, the four submodels are solved sequentially at each 10 s time step.

The hydrodynamic and solute transport submodels are described by a one-dimensional (1-D) representation. The river is discretised into subsections of variable lengths and the resolution of these two submodels requires the resolution of the whole river submodel. The erosion-sedimentation and sorption–desorption submodels, in turn, are not discretised. These submodels are applied to each river subsection without any interaction with the rest of the river.

2.1.1. Hydrodynamic submodel

The hydrodynamic model integrates 1-D Saint Venant equations, also known as the dynamic wave by Chow et al. (1988), to propagate the water in the river Eqs. (1) and (2):

\[ \frac{\partial Q}{\partial t} + \frac{\partial A}{\partial x} = 0 \]  

Fig. 1. Schematic representation of the TM modelling approach over a finite volume element of the river. The hydrodynamics and solute transport through advection–dispersion are represented in blue. They compute the longitudinal transport of both water and suspended elements (dissolved and particulate, blue arrows). The erosion-sedimentation model corresponds to the vertical transport of suspended particulate matter (SPM) between the immobile bottom of the river and the mobile water body. Finally, the trace metal model assesses the form taken by the trace metal (TM), either dissolved or particulate, through a sorption–desorption model based on a fixed separation coefficient Kd. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)
\[
\frac{\delta Q}{\delta t} + \frac{\delta (Q^2)}{\delta x} + ga \frac{\delta y}{\delta x} - S_h + \frac{Q^2}{R_h^2 A^2} = 0 \tag{2}
\]

where \(Q\) is the flow (\(m^3 \cdot s^{-1}\)), \(x\) the longitudinal distance along the channel (m), \(A\) the cross-section of the river (\(m^2\)), \(t\) the time (s), \(g\) the gravitational acceleration (\(m^2 \cdot s^{-2}\)), \(y\) the water level of the river (m), \(S_h\) the riverbed slope (\(m^{-1}\)), \(K_d\) the Strickler coefficient (\(m^{1.3} \cdot s^{-1}\)), inverse of the Manning coefficient) and \(R_h\) the hydraulic radius (m).

The Saint Venant equations are implemented in FORTRAN and solved using an implicit finite elements method. This implementation was first used in Sauvage et al. (2003). It takes as inputs the river slope, the bathymetry and boundary conditions (upstream and downstream) and provides as outputs the water height and velocity, the wet perimeter, the cross-section area of the river and some key hydraulic variables such as hydraulic radius and shear stress at the bottom of the river. It allows both steady state computations and dynamic simulations with variable inflow. The only free parameter is the Strickler coefficient.

2.1.2. Dissolved transport submodel

The transport model chosen is the one-dimensional transport with inflow and storage (OTIS) model developed by Bencala and Walters (1983). OTIS includes standard advection and dispersion equations and adds a transient storage reservoir that allows the heterogeneity of the river to be taken into account. The concentration of a given species in the river varies according to Eq. (3) and the concentration in the transient storage zone varies according to Eq. (4):

\[
\frac{\delta C}{\delta t} = \frac{Q \delta C}{A \delta x} + \frac{\delta}{\delta x} \left( \frac{AD \delta C}{\delta x} \right) + \alpha(C_s - C) \tag{3}
\]

\[
\frac{dC_s}{dt} = \frac{A}{A_s}(C - C_s) \tag{4}
\]

The additional variables are \(C\) the concentration of the transported species in the river (\(g \cdot m^{-3}\)), \(D\) the dispersion coefficient (\(m^2 \cdot s^{-1}\)) and \(C_s\) the concentration of the transported species in the transient storage zone (\(g \cdot m^{-3}\)). The OTIS model has two free parameters: the cross-section of the transient storage zone \(A_s\) (\(m^2\)) and the rate of exchange between the river and the transient storage \(\alpha\) (dimensionless). The governing flow \(Q\) and river cross-section \(A\) are provided by the hydrodynamic model. The dispersion coefficient can be calibrated, but can also be hydrodynamic conditions by the Eqs. (5) and (6):

\[
D = dR_h u^* \tag{5}
\]

\[
u^* = \frac{U \sqrt{g}}{K_s \times R_h^{1.5}} \tag{6}
\]

where \(d\) is a calibration parameter (unitless), \(u^*\) is the shear velocity (\(m \cdot s^{-1}\)) and \(U\) is the water velocity (\(m \cdot s^{-1}\)) computed by the hydrodynamic model.

2.1.3. Erosion and sedimentation submodel

Most erosion models use the equations of Partheniades (1965) to simulate the erosion and sedimentation of suspended sediments. These equations take advantage of the shear stress, a physically meaningful variable, at the bottom of the river to compute erosion or sedimentation. The sedimentation or erosion flux is computed using Eq. (7), based on the equations of Rossi (2004):

\[
\frac{dM}{dt} = \left\{ \begin{array}{ll}
C_{SPM} \times \text{We} \times \left( 1 - \frac{t}{t_{\text{ero}}} \right) & \text{if } t < t_{\text{ero}} \\
\frac{t}{t_{\text{ero}}} - 1 & \text{if } t > t_{\text{ero}}
\end{array} \right. \tag{7}
\]

where \(dM/dt\) is the eroded or deposited flux of SPM (\(g \cdot m^{-2} \cdot s^{-1}\)), \(C_{SPM}\) the concentration of SPM (\(g \cdot m^{-3}\)), \(\text{We}\) the sedimentation velocity (\(m \cdot s^{-1}\)), \(t_{\text{ero}}\) the critical shear stress for erosion \((Pa)\), \(t_{\text{eros}}\) the critical shear stress for sedimentation \((Pa)\) and \(\varepsilon\) the erosion constant (\(g \cdot m^{-2} \cdot s^{-1}\)).

Erosion is computed with respect to the hydrodynamic conditions using Eq. (8):

\[
\tau = \rho \times g \times \frac{U^2}{K_s R_h^{1.5}} \tag{8}
\]

where \(\rho\) is the density of water (\(kg \cdot m^{-3}\)).

The inputs to the erosion model are the shear stress at the bottom of the river \((\tau)\) and the concentration of SPM transported by the transport model. The model then needs four parameters to be calibrated: \(\text{We}, \varepsilon, t_{\text{ero}}\) and \(t_{\text{eros}}\). Particle TM is subject to the same equations and parameters as SPM. The only difference is the erosion constant, which is calculated using Eq. (9):

\[
e_{\text{TM}} = C_{\text{TM}} \times \varepsilon \tag{9}
\]

where \(e_{\text{TM}}\) is the erosion constant of the TM (\(g \cdot m^{-2} \cdot s^{-1}\)), \(C_{\text{TM}}\) is the concentration of TM in the SPM (gram of TM per gram of SPM) which needs to be measured or calibrated and \(\varepsilon\) is the same erosion constant of the SPM as in Eq. (7).

2.1.4. Sorption and desorption submodel

Trace metal concentrations are present in the water in two phases: particulate and dissolved. Both phases travel in the river continuum with the transport model. However, particulate elements can sediment in the pools with the help of the erosion-sedimentation model. In addition, freshly eroded SPM can contribute to the TM mass balance.

The sorption model chosen considers a constant \(K_d\), the partition coefficient between the dissolved and particulate phase based on SPM concentration. Since the erosion-sedimentation process represents a source and a sink of SPM, and thus of particulate TM, the equilibrium between the two phases of TM can be disrupted. The sorption model is then applied to re-equilibrate the two phases. The chosen form is Eq. (10), based on Allison and Allison (2005):

\[
K_d = \frac{\text{TM}_{\text{part}}(\mu g / kg)}{\text{TM}_{\text{diss}}(\mu g / l)} \tag{10}
\]

where \(K_d\) is the partition coefficient (\(kg^{-1}\)), \(\text{TM}_{\text{part}}\) the particulate concentration of TM in SPM (\(\mu g \cdot kg^{-1}\)) and \(\text{TM}_{\text{diss}}\) the dissolved concentration of TM (\(\mu g \cdot l^{-1}\)).

2.2. Study site

The Garonne River (southwest France) is the third largest river in France and is eighth order at its mouth. It exhibits a nivo-pluvial regime with the influence of winter snow precipitation in the Pyrenean mountains at its inlet (Pardé, 1935). Fig. 2a shows the extent of the Garonne River from the southern part of France to its outlet in the west. The study site is the middle course of the Garonne along an 80 km reach starting at the outlet of the city of Toulouse (800,000 inhabitants) (Fig. 2b). The downstream limit of the study site is before the influence of the Malause reservoir. The Garonne has a watershed of 13,730 km² and an annual average flow of 193 m³·s⁻¹. The monthly average flow...
ranges from 75 m$^3$ s$^{-1}$ in August to 341 m$^3$ s$^{-1}$ in May. The 10 day low-water average with a return period of two years is 42 m$^3$ s$^{-1}$ while the daily average flow of the two-year return flood reaches 1400 m$^3$ s$^{-1}$ (Banque Hydro, http://www.hydro.eaufrance.fr/).

The Garonne River exhibits successions of riffles and pools (Steiger and Gurnell, 2003). This morphological characteristic means that successive erosion and sedimentation zones can be observed in the 80 km reach of the study. In addition, an important gravel bed can be observed. Gravel beds are known to generate important interactions between the river and the groundwater for phosphorus and nitrate (Bonvallet Garay et al., 2001; Teissier et al., 2008). The Garonne watershed also exhibits a limited contamination in TM. Bur et al. (2009) reported enrichment factors for Pb up to 10.1 with respect to the regional geochemical background. Arsenic (As) levels are close to the geochemical background (Aubert et al., 2014; Drever, 1997), although the As geochemical background of the Garonne watershed is higher than the world average (Masson et al., 2007). Therefore, the Garonne River is subject to low-to-moderate metallic pollution along the study site.

2.3. Available data for model construction and calibration procedure

Different datasets were used to calibrate the submodel parameters sequentially. The automatic calibration was performed with the simplex method from Nelder and Mead (1965), implemented in the Octave free programming environment (http://www.gnu.org/software/octave/). The maximization of the Nash–Sutcliffe efficiency criterion (see Section 2.4) was used as the optimisation objective.

2.3.1. Hydrodynamic data

The morphology of the river is assessed by more than 100 cross-sectional profiles along the reach gathered by various public institutions such as France’s “Direction Régionale de l’Environnement” (DIREN – Regional Department for the Environment). These profiles allowed a detailed description of the riverbed slope and cross-section to be included in the model and were first used in Sauvage et al. (2003).

The Strickler parameter was calibrated and validated with hourly measurements of water levels and discharges from two gauging stations from 07/02/2002 to 08/04/2002. This period showed flows ranging from low to medium, thus allowing simulations over a significant range of water velocity. The water levels were measured and converted into discharges by the Midi-Pyrenees “Direction Régionale de l’Environnement, de l’Aménagement et du Logement” (DREAL – Regional Directorate for Environment, Development and Housing). The first gauging station is situated at Portet, upstream of Toulouse (see Fig. 2), while the second is at Verdun (point G5 on Fig. 2), 49.5 km downstream in the middle part of the studied sector.

2.3.2. Dissolved transport model – Tracer tests

Two rhodamine B tracer test campaigns were performed on the Garonne River in 2007 and 2008 by private companies E.A.T.C. and SETUDE, 2008. The campaigns aimed to assess the dispersion of a dissolved pollutant in the Garonne under various hydrological conditions. Two hydrological conditions were considered, namely low water (61 m$^3$ s$^{-1}$) and high water (447 m$^3$ s$^{-1}$). As the studied sector was too long for a single injection, the river was separated into three sections of 40 km, 20 km and 10 km, with the first injection performed upstream of Toulouse, the second at point G4 and the third half-way between points G5 and G6 (Fig. 2). The samples were taken by an automated sampler and kept in darkness to prevent photo-degradation. The rhodamine B analyses were performed using a spectrophotometer.

When using the 1-D OTIS model, it is important that the injected tracer is fully mixed with the river. The tracer was considered fully mixed in the river when the sampling from the middle of the river showed a similar behavior to the samples taken from the two river banks. The first sampling point meeting this
criterion was chosen as the injection point for the simulation of the tracer test. Two injections were used to calibrate the model. The first was performed on 19 September 2007, during a low-water flow period of 61 m$^3$s$^{-1}$. The injection was performed 5 km downstream of point G1 and the observations were made at point G4, for a total distance of 28.1 km. The second injection was performed on 1 April 2008, during a flood event with a discharge of 447 m$^3$s$^{-1}$. The injection was made at point G2 and the observations were made at point G3, for a total distance of 12.7 km. These observations allowed a calibration of the three parameters of the OTIS model, namely the storage cross-section area AS, the exchange coefficient α and the dispersion parameter d. This last parameter d was set to 100 based on previous results (Sauvage et al., 2003). The two remaining parameters were estimated by automatic calibration for both low water and flood events.

2.3.3. Suspended particulate matter and trace metal data

SPM and trace metals have been measured in the waters of the Garonne River and its tributaries (Sánchez-Pérez et al., 2006) in the various sampling stations on the Garonne River (Gx) and its tributaries (Tx) indicated in Fig. 2, according to the sampling method of Teissier et al. (2008). The sampling method consisted of sampling the same water body as it progressed to the outlet of the river. Therefore, when a flood event occurred, the various sampling points were sampled by following the flood wave to the outlet of the river. The first campaign was performed on 17 and 18 May 2005 during a flood event (flow of 800 m$^3$s$^{-1}$). The second occurred on 10 November 2005 after three weeks of low water (flow of 80 m$^3$s$^{-1}$) (Fig. 3). SPM from upstream of the river and from the tributaries measured during the two sampling campaigns were used as inputs for the erosion-sedimentation submodel. Observations along the longitudinal profile were used for the calibration of the four free parameters: $T_{e0x}$, $T_{e0d}$, $W_r$ and $e$.

Within the scope of this work, two trace metals – arsenic (As) and lead (Pb) – were chosen as they are known to have a different chemical affinity to suspended matter in the Garonne River and thus different sorption/desorption behaviors (Allison and Allison, 2005). Arsenic mainly exhibits a dissolved fraction under most hydrological conditions with log(Kd) between 2.0 and 6.0 and a median of 4.0 according to Allison and Allison (2005). In contrast, lead is usually encountered in particulate form. Log(Kd) found in literature ranges from 3.4 to 6.5 with a median value of 5.6 (Allison and Allison, 2005). This data is consistent with data observed at the Portet station (Aubert et al., 2014). The samples were treated according to the procedures described by in N’guessan et al. (2009), Roussiez et al. (2013) and Aubert et al. (2014) and summarized below. Water samples were filtrated through 0.22 μm Millipore GSWP-type membranes for trace metal analysis. SPM was removed from the filter using ultrasound with a recovery rate of between 70 and 100% to avoid filter contamination for some elements that can occur if there is complete dissolution. Particles were then transferred into Teflon bombs and evaporated until completely dry. SPM dissolution was performed in a hot mixture of bi-distilled HNO$_3$ and HF and ultrapure H$_2$O$_2$ in a clean room using a well-calibrated procedure. Trace elements (As and Pb) in the dissolved fraction of the water and SPM were measured using a PerkinElmer Elan 6000 ICP-MS. The detection limit was below 0.01 μg/l and the error of measurement was generally less than 5% for most of the elements. Standard materials were used during the procedure allowing effective control of the dissolution procedure and the analytical process.

Two model parameters are specifically related to the TM simulations: Kd and the concentration of TM in the SPM $C_{TM}$.

Within the scope of the study, the Kd coefficient was fixed to the average Kd observed during the sampling campaigns. This choice allowed one free parameter to be removed from the model’s auto-calibration.

2.4. Efficiency criteria

Two efficiency criteria were used to assess the performance of the calibrations. The first was the Nash–Sutcliffe criterion (NS, Eq. (11)), which compared the square error between the observation and the simulation to the square error between the observation and the mean of the observations. An NS of 1 is indicative of a perfect model, a value of 0 means that the model is a predictor that is as good as the average of the observations, and a value below 0 means that the average of the observations is a better predictor than the model.

$$ NS = 1 - \frac{\sum(Q_{obs} - Q_{sim})^2}{\sum(Q_{obs} - \bar{Q}_{obs})^2} $$

Eq. (11) was used to assess the performance of the hydrodynamic model, the transport module and the erosion module. However, the trace metal module generates three components (dissolved phase, particulate phase and total concentration) and a single criterion was needed to perform the calibration. Since the average of the three components has no physical meaning, the Nash–Sutcliffe was modified (Eq. (12)). This modification allowed the components of different magnitudes to be compared:

$$ NS_{mod} = 1 - \frac{\sum(C_{diss,obs} - C_{diss,sim})^2 + \sum(C_{parbs,obs} - C_{parbs,sim})^2}{\sum(C_{diss,obs} - \bar{C}_{obs})^2 + \sum(C_{parbs,obs} - \bar{C}_{obs})^2 + \sum(C_{totbs,obs} - \bar{C}_{obs})^2} $$

(12)

The second efficiency used was the coefficient of correlation that quantifies the correlation between two variables (Eq. (13)). For trace metals, the average of the correlation of the dissolved phase, the particulate phase and the total concentration was provided:

$$ r = \frac{\sigma_{Q_{obs}Q_{sim}}}{\sigma_{Q_{obs}} \times \sigma_{Q_{sim}}} $$

(13)

3. Results and discussion

3.1. Hydrodynamic model

River discharge was calibrated with a mean Strickler coefficient of 48 (Fig. 4). The NS of the calibration period (12 days of hourly observations) is 0.82 while the r is 0.98, which means that the model is capable of representing the flow well. For the 10-day validation period, the NS reaches 0.96 and the r is 0.97.

The hydrodynamic model demonstrated a good representation of the observations. However, over the 49.5 km between the two

![Fig. 3. Discharge variations of the Garonne River at station G5 during 2005. SPM and TM sampling campaigns are indicated by the grey line.](image-url)
gauging stations, the Strickler parameter only influenced the velocity of the water. The mismatch between the observed flow and simulated flow includes the mass balance error (unknown inputs from smaller tributaries or uptake by irrigation) and measures uncertainty.

A Strickler parameter of 48 is similar to a Manning coefficient of 0.021. According to literature (Chow et al., 1988), a Manning coefficient of this level is characteristic of rivers that have a gravel bed and concrete or mortared stone sides. This description fits the Garonne River well as it exhibits a gravel bed with frequent ripraps along the studied sector (Steiger et al., 1998). The one-dimensional model provided a good relationship between the automatic calibration and physical signification of the Manning parameter thanks to the large number of cross-section profiles providing a very good bathymetry of the river.

3.2. Transport model

The tracer experiment allowed the OTIS model to be calibrated and the influence of the morphology of the river on the transport of dissolved element to be quantified. Furthermore, calibrating the hydraulics first allowed the velocity of the flow to be distinguished from the velocity of the dissolved elements in the water. The morphology of the river could therefore be highlighted.

The transport model was calibrated on two distinct tracer campaigns: low and high water (61 m$^3$/s$^1$ and 447 m$^3$/s$^1$, Fig. 5). The optimal parameters are shown in Table 1. The dispersion parameter $d$ of Eq. (5) was fixed to 100 based on the results of Sauvage et al. (2003) in the same sector.

The optimal parameter sets were different for low water and high water, although they were of the same magnitude in both hydrological conditions. The transient storage cross-section was greater during high water, which was to be expected as the river cross-section is larger during high water than during low water. The exchange rate, however, reacted inversely by having a higher value during low water. The dispersion range observed was of the same order of magnitude for both events. Using two different parameter sets allowed NS and $r$ criterion very close to one to be obtained, which indicated a very good fit. The parameters compared favorably with literature (Bencala and Walters, 1983; Ambrose et al., 1993; Runkel et al., 1999). Furthermore, the application of OTIS on the Garonne River case study was strengthened by the morphology of the river, which shows a succession of riffles and pools (Pardé, 1935; Sauvage et al., 2003), since Runkel et al. (1999) suggested that the pools can have a significant storage capacity of dissolved elements capable of slowing their transport. Peyrard et al. (2008) also showed the activity of the hyporheic zone in the studied sector, thus increasing the potential volume of the transient storage zone.

Accurate simulation of the transport of the dissolved element required consideration of the river’s transient storage zones. In fact, by ignoring transient storage, the only free parameter is $d$ and its optimal value was 4,865 during high water and 17,544 during low water. Such values lead to a dispersion coefficient $D$ in the range of 72–1370 m$^2$/s$^1$ during high water and 5–5,260 m$^2$/s$^1$ during low water and gave an NS criterion of 0.63 (high water) and 1.6 (low water). These unrealistic dispersion coefficients also resulted in the simulated tracer peak concentrations occurring...
The calibrated optimal parameter set also suggested that the nature of SPM varied with hydrological conditions. The settling velocity $W_c$ provided information on the smallest particles that could settle. In accordance with Stokes’ law, it was possible to estimate the diameter of a particle with a specific settling velocity by using Eq. (14):

$$W_c = \frac{2(\rho_{water} - \rho_{SPM})gR^2}{\mu}$$

(14)

where $\rho$ is the density of water and SPM, respectively (kg m$^{-3}$), $g$ is the gravitational acceleration (m s$^{-2}$), $\mu$ is the dynamic viscosity (Pa s) of water and $R$ is the average radius of the SPM (m).

By assuming an average SPM density of 2700 kg m$^{-3}$, the diameter of the smallest particles that could settle was around 20 µm during low-water periods and 80 µm during high water. Since fine sands are usually associated with SPM having a diameter greater than 60 µm, the computed settling velocities suggested that fine sands could be transported in SPM during high waters, but not during low-water periods. These values were consistent with the observations of Steiger and Gurnell (2003) who studied sediments deposited after three flood events and found that particles greater than 63 µm contributed up to 85% of the total deposited sediment.

The results from the different hydrological conditions also suggested that a dynamic model capable of simulating the fate of SPM under all hydrological conditions would need at least two classes of SPM. The first would include the very fine material that is likely to stay in suspension along the whole river and the second would represent the fine material capable of sedimentation.

### 3.3. Erosion model

The calibrated results and parameters of the SPM concentrations are shown in Fig. 6 and Table 2, respectively for low and high-flow conditions. The NS of the low-flow simulation was 0.04 but the correlation coefficient $r$ of 0.87 ($n = 7$, $p < 0.01$) indicated that the general trend of the SPM concentration was well represented by the model. This low NS is typical of simulation results close to their average value (Cris and Winston, 2008) and has to be interpreted with the $r$ coefficient.

During the high-water flow event, the model performed much better with NS of 0.45 and $r$ of 0.96 ($n = 7$, $p < 0.01$). The comparison of the low-flow and high-flow parameter sets and simulation trends suggested that SPM characteristics might be different during the two hydrologically contrasting periods. It should also be noted that the concentrations of SPM were almost two orders of magnitude higher during the flood period than during the low flow (Fig. 6).

The critical shear stress of sedimentation ($\tau_{sed}$) and the falling velocity ($W_c$) exhibited more than one order of magnitude of difference between the two events, being much lower during the low-water event. The parameters related to erosion (the critical shear stress of erosion $\tau_{eros}$ and the erosion constant $e$) were much higher during the flood event, but remained within one order of magnitude of the values calculated during the low waters (Table 2).

The most plausible explanation for the mismatch between the model and the observations is the absence of autochthonous organic matter production in the model. The Garonne River is known to have significant epilithic biofilm production during low-water flows and many attempts have been made to simulate its fate (Boulêtreau et al., 2006; Graba et al., 2013). This biofilm produces particulate organic carbon that can contribute up to 25% of SPM during low-water periods (Veyssy et al., 1998).

During high water, the simulation results were better, suggesting that transport and sedimentation were well represented while erosion had a limited influence. The low erosion contribution suggested that SPM did not originate from the river bed, but mostly from the watershed.

The calibrated optimal parameter set also suggested that the nature of SPM varied with hydrological conditions. The settling velocity $W_c$ provided information on the smallest particles that could settle. In accordance with Stokes’ law, it was possible to estimate the diameter of a particle with a specific settling velocity by using Eq. (14):

$$W_c = \frac{2(\rho_{water} - \rho_{SPM})gR^2}{\mu}$$

(14)

where $\rho$ is the density of water and SPM, respectively (kg m$^{-3}$), $g$ is the gravitational acceleration (m s$^{-2}$), $\mu$ is the dynamic viscosity (Pa s) of water and $R$ is the average radius of the SPM (m).

By assuming an average SPM density of 2700 kg m$^{-3}$, the diameter of the smallest particles that could settle was around 20 µm during low-water periods and 80 µm during high water. Since fine sands are usually associated with SPM having a diameter greater than 60 µm, the computed settling velocities suggested that fine sands could be transported in SPM during high waters, but not during low-water periods. These values were consistent with the observations of Steiger and Gurnell (2003) who studied sediments deposited after three flood events and found that particles greater than 63 µm contributed up to 85% of the total deposited sediment.

The results from the different hydrological conditions also suggested that a dynamic model capable of simulating the fate of SPM under all hydrological conditions would need at least two classes of SPM. The first would include the very fine material that is likely to stay in suspension along the whole river and the second would represent the fine material capable of sedimentation.

### 3.4. Trace metal model

The longitudinal profiles of observed and simulated concentrations for As and Pb are shown in Fig. 7 and the calibrated parameters and efficiency criterion in Table 3. The parameters used for both trace metals showed little variability independent of the hydrological condition. The $NS_{sim}$ score indicated a better agreement between observations and simulation during high-water flow conditions for both TM, while the average $r$ suggested that the model could simulate the trend of the two TM.

During the low-water flow period, As was highly dissolved (Fig. 7A), as has already been observed in the Garonne River (Aubert et al., 2014). Nevertheless, during high-water flows, the much higher concentration of SPM led to a larger fraction of As in the particulate phase than in the dissolved one (Fig. 7B). The increase of concentration of total As could also be linked almost completely to the particulate fraction of As, since the dissolved fraction remained in similar concentrations during both events. Therefore, although As is quite soluble, in the presence of high SPM

![Fig. 6. SPM concentration simulated along the river continuum during low water (A, flow of 80 m$^3$ s$^{-1}$, continuous line) and during high water (B, flow of 800 m$^3$ s$^{-1}$, continuous line). Observation values at the sampling points (crosses, see Fig. 2) showed a different scale in the two hydrological conditions.](image-url)
concentrations the particulate form dominated, and erosion and sedimentation were the most important processes for explaining the fate of As.

The final three observations (between kilometers 740 and 760) of lead during the low-water simulation revealed significant concentrations (Fig. 7C). These points made the use of auto-calibration problematic, thus a manual calibration was performed. Nevertheless, the simulation of Pb in low-water conditions followed the trend of the observations with an average $r$ of 0.44, although $NS_{mod}$ was lower (0.34, Fig. 7C). Similar to SPM and As dynamics, during high-water conditions (Fig. 7D), total concentrations of lead were over one order of magnitude higher than for low waters. As with the arsenic simulations, the dissolved fraction of lead was predicted effectively in both low and high waters and corresponded to the observations. This highlighted the fact that the increase in total concentration was due to the particulate fraction of lead and also that the particulate fractions of both trace metals closely follow the trend of SPM concentration.

The variation in $K_d$ can be explained by a variation in the physicochemistry of the water as well as by a variation in the nature of the SPM between contrasted hydrological periods. Such dependency of $K_d$ with respect to physicochemistry and SPM composition during contrasted hydrological conditions is well known (Aubert et al., 2014; Allison et al., 1991; Tipping, 1994; Parkhurst and Appelo, 1999; Roussiez et al., 2013).

The concentration of TM in the bed sediment was fixed to obtain the best fit with the observations. If no sediment analysis was performed during the sampling campaign, SIEAG (Système d’Information sur l’Eau du Bassin Adour-Garonne – Information centre on water from the Adour-Garonne watershed, http://adour-garonne.eaufrance.fr/accesData/) collected and analysed the sediment over the 2003–2011 period, with one sample taken once a year during the low-flow period at sampling point G5 (Fig. 2). Lead concentration ranged from 5 to 23 mg$g^{-1}$ of sediment with a value of 10 mg$g^{-1}$ for 2005, the year of the sampling campaign for the present study. Arsenic concentrations ranged from 2.9 to 13 mg$g^{-1}$ with a value of 8 mg$g^{-1}$ for 2005. For both trace metals, the calibrated concentration of TM in the sediment was in the range of the concentrations observed by SIEAG. However, these concentrations are site-specific (Aubert et al., 2014; N’guessan et al., 2009). N’guessan et al. (2009) sampled the sediment on the left tributaries of the Garonne watershed from the molassic area and found average lead concentrations of 24.90 mg$g^{-1}$ and arsenic concentrations of 15.14 mg$g^{-1}$. If these values are of the same

<table>
<thead>
<tr>
<th>Table 3</th>
<th>Calibrated values of the trace metal model parameters for the two hydrological conditions considered (low and high waters).</th>
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</thead>
<tbody>
<tr>
<td></td>
<td>Log($K_d$)</td>
</tr>
<tr>
<td>As low water</td>
<td>4.15</td>
</tr>
<tr>
<td>As high water</td>
<td>4.28</td>
</tr>
<tr>
<td>Pb low water</td>
<td>6.33</td>
</tr>
<tr>
<td>Pb high water</td>
<td>6.55</td>
</tr>
</tbody>
</table>
magnitude as the calibrated parameters, the $C_{TM}$ parameter has been found to be too sensitive to use these observations directly. Nevertheless, the calibrated value of TM concentration in the bed sediment proved capable of providing information about TM contamination in the sediment.

In all four cases (Fig. 7A–D), the simulation did not match the final three sampling points (G6, G7 and G8). This was the case in particular for Pb during low-water conditions. Considering the significant positive correlation between the trace metals and SPM, it can be hypothesised that improving the erosion model would significantly improve the TM model. This sector also showed the lowest density of river profiles, with sectors up to 2.4 km without profile. This low resolution of the bathymetry could lead to a poor description of the riffle–pool sequence, thus leaving out high sedimentation and erosion zones.

### 4. Conclusion

The modelling of trace metals in rivers is heavily dependent on hydrodynamics, the transport of SPM and the separation between the dissolved and particulate phases. In the present case, a dynamic, discretised model of trace metals is presented. It includes a detailed description of the hydrodynamics comprising a dynamic, discretised model of trace metals is presented. It includes hydrodynamics, the transport of SPM and the separation between SPM and TM, and separation of TM into dissolved and particulate phases. The model presented required the calibration of ten parameters and providing insight into these properties. The strategy of using different datasets highlighted: (a) the importance of transient storage in the study case, (b) a detailed description of the erosion and sedimentation processes of SPM, and (c) the significance of erosion as a source of TM for surface water.

The proposed model also highlighted the influence of hydrology on the composition of SPM, and thus on the fate of TM, since the parameters were different between the two hydrological conditions. It also suggests that at least two classes of SPM should be included in the model in order to produce a model capable of tackling all hydrological conditions. Furthermore, these classes could represent different origins of SPM or its composition (i.e. organic matter from autochthonous or allochthonous origin). Finally, a detailed description of the physicochemistry of the water to assess the sorption–desorption equilibrium of TM through complex chemical models should be investigated in order to quantify its importance in the transfer of TM on larger temporal and spatial scales.

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### References


Lindenschmidt, K., Poser, K., Rode, M., 2005. Impact of morphological parameters on the composition of SPM, and thus on the fate of TM, since the parameters were different between the two hydrological conditions. Furthermore, these classes could represent different origins of SPM or its composition (i.e. organic matter from autochthonous or allochthonous origin). Finally, a detailed description of the physicochemistry of the water to assess the sorption–desorption equilibrium of TM through complex chemical models should be investigated in order to quantify its importance in the transfer of TM on larger temporal and spatial scales.

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