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Modeling inorganic carbon dynamics in the Seine River continuum in France

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Abstract

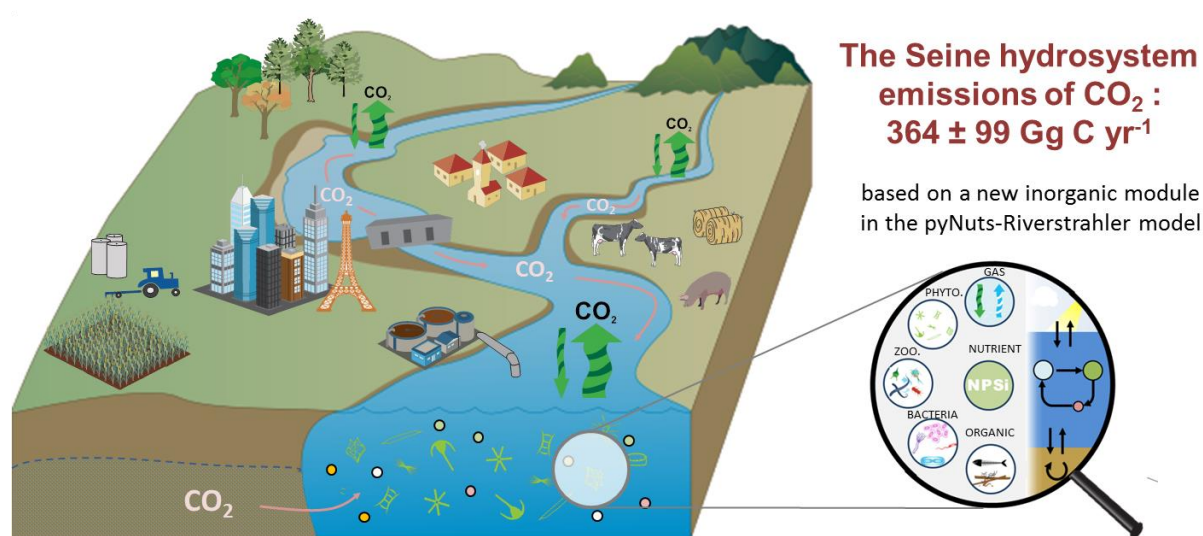
Inland waters are an active component of the carbon cycle where transformations and transports are associated with carbon dioxide (CO₂) outgassing. This study estimated CO₂ emissions from the human-impacted Seine River (France) and provided a detailed budget of aquatic carbon transfers for organic and inorganic forms, including in-stream metabolism along the whole Seine River network. The existing process-based biogeochemical pyNuts-Riverstrahler model was supplemented with a newly developed inorganic carbon module and simulations were performed for the recent time period 2010-2013. New input constraints for the modelling of riverine inorganic carbon were documented by field measurements and complemented by analysis of existing databases. The resulting dissolved inorganic carbon (DIC) concentrations in the Seine aquifers ranged from 25 to 92 mgC L⁻¹, while in wastewater treatment plant (WWTP) effluents our DIC measurements averaged 70 mgC L⁻¹.

Along the main stem of the Seine River, simulations of DIC, total alkalinity, pH, and CO₂ concentrations were of the same order of magnitude as the observations, but seasonal variability was not always well reproduced. Our simulations demonstrated the CO₂ supersaturation with respect to atmospheric concentrations over the entire Seine River network. The most significant outgassing was in lower order streams while peaks were simulated downstream of the major WWTP effluent. For the period studied (2010–2013), the annual average of simulated CO₂ emissions from the Seine drainage network were estimated at 364 ± 99 Gg C yr⁻¹.

Results from metabolism analysis in the Seine hydrographic network highlighted the importance of benthic activities in headwaters while planktonic activities occurred mainly downstream in larger rivers. The net ecosystem productivity remained negative throughout the 4 simulated years and over the entire drainage network, highlighting the heterotrophy of the basin.

Keywords: CO₂ outgassing; inorganic carbon modeling; instream metabolisms; waste-
and ground water inputs; carbon budget ; temperate Seine River

Graphical abstract:



Highlights:

- CO₂ emission from the Seine River was estimated at 364 ± 99 GgC yr⁻¹ with the Riverstrahler model.
- CO₂ riverine concentrations are modulated by groundwater discharge and instream metabolism.
- CO₂ emissions account for 31% of inorganic carbon exports, the rest being exported as DIC.

1. Introduction

Rivers have been demonstrated to be active pipes for transport, transformation, storage and outgassing of inorganic and organic carbon (Cole et al., 2007). Although there are large uncertainties in the quantification of flux from inland waters, carbon dioxide (CO₂) outgassing has been estimated to be a significant efflux to the atmosphere, subjected to regional variabilities (Cole et al., 2007; Battin et al., 2009a; Aufdenkampe et al., 2011; Lauerwald et al., 2015; Regnier et al., 2013; Raymond et al., 2013a; Sawakuchi et al., 2017; Drake et al., 2017). These variabilities are determined by regional climate and watershed characteristics and are related to terrestrial carbon exports under different forms, from organic to inorganic, and dissolved to particulate. Organic carbon entering rivers can originate from terrestrial ecosystems as plant detritus, soil leaching or soil erosion and groundwater supply, but it can also be produced instream by photosynthesis or brought by dust particles (Prairie and Cole, 2009; Drake et al., 2017). Inorganic carbon originate from groundwater, soil leaching and exchange by diffusion at the air–water interface, depending on the partial pressure of CO₂ (pCO₂) at the water surface with respect to atmospheric pCO₂ (Cole et al., 2007; Drake et al., 2017; Marx et al., 2018). Beside air-water exchanges, carbon exchanges occur at the water–sediment interface, through biomineralization and/or burial (Regnier et al., 2013b). As a whole, eutrophic, oligo- and mesotrophic hydrosystems generally act as a source of carbon however, lentic systems may be undersaturated with respect to atmospheric pCO₂ (Prairie and Cole, 2009; Xu et al., 2019; Yang et al., 2019).

Direct measurements of pCO₂ or isotopic surveys (as realized by Dubois et al. 2010 in the Mississippi River) along the drainage network are still too scarce to accurately support temporal and spatial analyses of CO₂ variability. While calculations from pH, temperature and alkalinity may help reconstruct spatiotemporal patterns of CO₂ dynamics (Marescaux et al., 2018b), modeling tools can predict the fate of carbon in whole aquatic systems. Indeed, modeling

approaches have made it possible to simulate and quantify carbon fluxes between different reservoirs: atmosphere, biosphere, hydrosphere and lithosphere (e.g., Bern-SAR, Joos et al., 1996; ACC2, Tanaka et al., 2007; TOTEM, Mackenzie et al., 2011; MAGICC6, Meehl et al., 2007). In addition to these box approaches, a number of more comprehensive mechanistic models, describing biogeochemical processes involved in carbon cycling and CO₂ evasion, have been set up for oceans (e.g., Doney et al., 2004; Aumont et al., 2015), coastal waters (e.g., Borges et al., 2006; Gypens et al., 2004, 2009, 2011) and estuaries (e.g., Cai and Wang, 1998; Volta et al., 2014, Laruelle et al., 2019). In inland waters, the NICE-BGC model (Nakayama, 2016) accurately represents CO₂ evasion at the global scale. However, to our knowledge, while several process-based river models describe the carbon cycle through organic matter input and degradation by aquatic microorganisms (e.g., PEGASE, Smits et al., 1997; ProSe, Vilmin et al., 2018; QUAL2Kw, Pelletier et al., 2006; QUAL-NET, Minaudo et al., 2018, QUASAR, Whitehead et al., 1997; Riverstrahler, Billen et al., 1994; Garnier et al., 2002), none of them describes the inorganic carbon cycle including carbon dioxide outgassing.

The Seine River (northwestern France) has long been studied using the biogeochemical riverine Riverstrahler model (Billen et al., 1994; Garnier et al., 1995), a generic model of water quality and biogeochemical functioning of large river systems. For example, the model has made it possible to quantify deliveries to the coastal zone and understand eutrophication phenomena (Billen and Garnier, 2000; Billen et al., 2001; Passy et al., 2016; Garnier et al., 2019), nitrogen transformation and N₂O emissions (Garnier et al., 2007, 2009; Vilain et al., 2012) as well as nitrate retention (Billen and Garnier, 2000; Billen et al., 2018), and organic carbon metabolism (Garnier and Billen, 2007; Vilmin et al., 2016). It is only recently that we investigated pCO₂ and emphasized the factors controlling pCO₂ dynamics in the Seine River (Marescaux et al., 2018b) or its estuary (Laruelle et al., 2019).

The purpose of the present study was to quantify the sources, transformations, sinks and gaseous emissions of inorganic carbon using the Riverstrahler modelling approach (Billen et al., 1994; Garnier et al., 2002; Thieu et al., 2009). A further aim in newly implementing this CO₂ module was to quantify and discuss autotrophy versus heterotrophy patterns in regard to CO₂ concentrations and supersaturation in the drainage network.

2. Material and methods

2.1. Description of the Seine basin

Situated in northwestern France, 46°57' – 50°55' north and 0°7' 1" – 4° east, the Seine basin (~76,285 km²) has a temperate climate and a pluvio-oceanic hydrologic regime (Figure 1). The mean altitude of the basin is 150 m above sea level (ASL) with 1% of the basin reaching more than 550 m ASL in the Morvan (Guerrini et al., 1998). The water flow at Poses (stream order 7, basin area 64,867 km²), the most downstream monitoring station free from tidal influence, averaged 490 m³ s⁻¹ during the 2010–2013 period (the HYDRO database, <http://www.hydro.eaufrance.fr>, last accessed 2020/02/11). The major tributaries include the Marne and upper Seine rivers upstream from Paris, and the Oise River downstream from Paris (Figure 1a). Three main reservoirs, storing water during winter and sustaining low flow during summer, are located upstream on the Marne River and the upstream Seine and its Aube tributary (Figure 1a). The total storage capacity of these reservoirs is 800 10⁶ m³ (Garnier et al., 1999).

The maximum water discharge of these tributaries occurs during winter with the lowest temperature and rate of evapotranspiration; the opposite behavior is observed during summer (Guerrini et al., 1998).

Except for the crystalline rocks in the north and from the highland of the Morvan (south), the Seine basin is for the most part located in the lowland Parisian basin with sedimentary rocks (Mégnyen, 1980; Pomerol and Feugueur, 1986; Guerrini et al., 1998). The largest aquifers are

125 in carbonate rock (mainly limestone and chalk) or detrital (sand and sandstone) material
126 separated by impermeable or less permeable layers.

127 The concept of Strahler stream order (SO) (Strahler, 1957) was adopted for describing the
128 geomorphology of a drainage network in the Riverstrahler model (Billen et al., 1994). The
129 smaller perennial streams are order 1. Only confluences between two river stretches with the
130 same SO produce an increase in Strahler ordination (SO+1) (Figure 1a). The mean
131 hydrophysical characteristics of the Seine River are aggregated by stream orders shown in Table
132 1.

133 The Seine basin is characterized by intensive agriculture (more than 50% of the basin, CLC -
134 EEA, 2012) is mostly concentrated in the Paris conurbation (12.4 million inhabitants in 2015)
135 (Figure 1) (INSEE, 2015). Located 70 km downstream of Paris, the largest wastewater
136 treatment plant in Europe (Seine Aval, SAV WWTP) can treat up to $6 \cdot 10^6$ inhab eq per day,
137 releasing $15.4 \text{ m}^3 \text{ s}^{-1}$ into the lower Seine River (*Syndicat interdépartemental pour*
138 *l'assainissement de l'agglomération parisienne*; French acronym SIAAP, <http://www.siaap.fr/>,
139 last accessed 2020/02/11).

SO	Draining area km^2	Cum. length Km	Width (*) m	Depth (**) m	Slope (*) m m^{-1}	Discharge (**) $\text{m}^3 \text{ s}^{-1}$	Flow velocity (**) m s^{-1}
1	36083	12759	2.4	0.14	0.01442	0.13	0.34
2	12354	5231	5.2	0.29	0.00540	0.66	0.36
3	7067	2871	10.6	0.45	0.00300	2.17	0.47

4	4054	1548	20.2	0.79	0.00212	6.35	0.33
5	2649	943	46.0	1.11	0.00060	25.87	0.46
6	2094	636	77.8	2.51	0.00029	82.22	0.42
7	1354	318	168.3	2.61	0.00037	416.16	0.81

Table 1: Hydro-morphological characteristics of the Seine drainage network, (*) averaged by Strahler order (SO) and (**) over the time period 2010-2013. Hydrographic network provided by the Agence de l'Eau Seine Normandie and water discharges by the national Banque Hydro database. Depth and flow velocity calculated according to Billen et al 1994; width calculated according to Thieu et al., 2009.

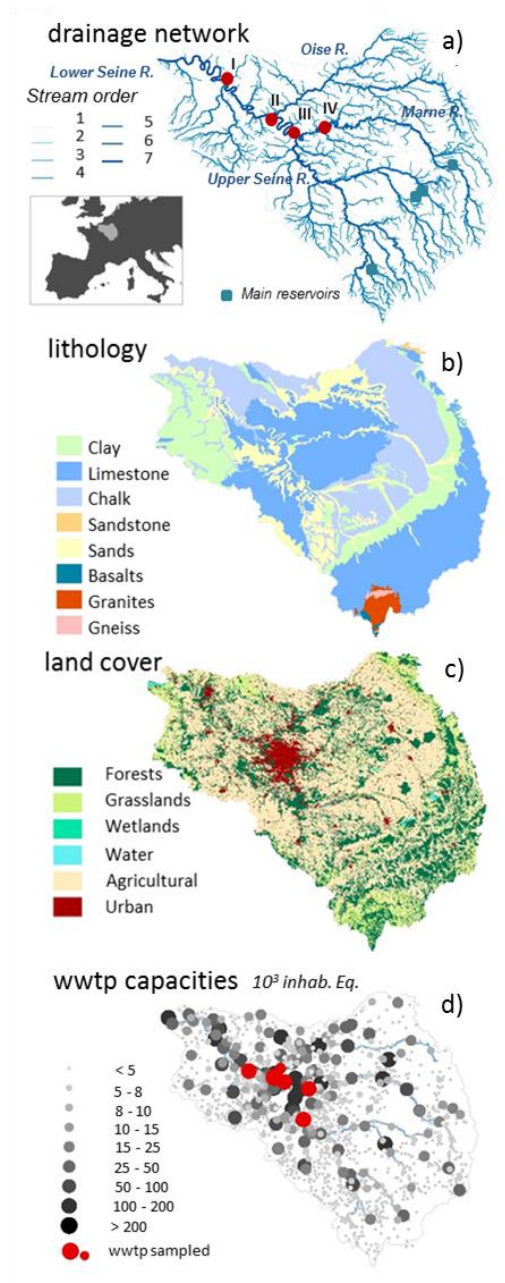


Figure 1 Characteristics of the Seine basin: a) drainage network according to Strahler stream orders (Strahler, 1952, 1957), monitoring stations (I: Poses, II: Poissy (Downstream Paris), III: Paris, IV: Ferté-sous-Jouarre (Upstream Paris); b) the lithology according to Albinet, (1967); c) land use according to the Corine Land Cover database, with six simplified classes (EEA, 2012); d) wastewater treatment plants (WWTPs) of the basin. Red dots are the WWTPs sampled in 2018.

2.2. The pyNuts-Riverstrahler model and its biogeochemical model, RIVE

The biogeochemical model, RIVE. The core of the biogeochemical calculation of the pyNuts-Riverstrahler model (described hereafter) is the RIVE model (e.g., Billen et al., 1994; Garnier et al., 1995; Garnier et al., 2002; Servais et al., 2007) (<https://www.fire.upmc.fr/rive/>), which simulates concentrations of oxygen, nutrients (nitrogen (N), phosphorus (P) and silica (Si)), particulate suspended matter, and dissolved and particulate organic carbon (three classes of biodegradability) in a homogeneous water column. Biological compartments are represented by three taxonomic classes of phytoplankton (diatoms, Chlorophyceae and Cyanobacteria), two types of zooplankton (rotifers with a short generation time and microcrustaceans with a long generation time), two types of heterotrophic bacteria (small autochthonous and large allochthonous with a higher growth rate than the small ones), as well as two types of nitrifying bacteria (ammonium-oxidizing bacteria and nitrite-oxidizing bacteria).

The model also describes benthic processes (erosion, organic matter degradation, denitrification, etc.) and exchanges with the water column with the explicit description of benthic organic matter, inorganic particulate P and benthic biogenic Si state variables. The benthic component does not explicitly represent all the anaerobic reduction chains, denitrification being the major anaerobic microbial process.

A detailed list of the state variables of the RIVE model is provided in S1. Most of the kinetic parameters involved in this description have been previously determined through field or laboratory experiments under controlled conditions and are fixed a priori (see detailed description of all kinetics and parameters values in Garnier et al., 2002). To date, there has been no explicit representation of inorganic carbon in the RIVE model (see this new input in S1).

Riverstrahler allows for the calculation of water quality variables at any point in the aquatic continuum based on a number of constraints characterizing the watershed, namely, the geomorphology and hydrology of the river system and the point and diffuse sources of nutrients.

Geomorphology. A drainage network can be described as subbasins (tributaries) connected to one or several main axes, that define a number of modelling units. The modelling approach considers the drainage network as a set of river axes with a spatial resolution of 1 km (axis-object), or they can be aggregated to form subbasins that are idealized as a regular scheme of tributary confluences where each stream order is described by mean characteristics (basin-object). Here, the Seine drainage network starts from headwater until its fluvial outlet (Poses) and was divided into 69 modeling units, including six axes (axis-object) and 63 upstream basins (basin-object). A map and a table introducing the main characteristics of the modeling units are provided in S2.

Hydrology. Runoffs were calculated over the whole Seine basin using water discharge measurements at 48 gauged stations (source: Banque Hydro database, <http://www.hydro.eaufrance.fr/>, last accessed 2020/02/11). Surface and base flow contributions were estimated applying the BFLOW automatic hydrograph separation method (Arnold and Allen, 1999) over the recent time series of water discharges (2010–2017). For the study period (2010–2013), the mean base flow index (BFI = 0.71) of the Seine basin indicates the extent of the groundwater contribution to river discharge, with spatial heterogeneity following the main

lithological structures (Figure 1b), but when summarizing the BFI criteria by Strahler order, significant differences did not appear (not shown).

Water temperature. Water temperature was calculated according to an empirical relationship, adjusted on inter-annual averaged observations (2006—2016), and describes seasonal variation of water temperature in each Strahler order with a 10-day time step (see S2).

Diffuse and point sources. Riverstrahler manages the calculation of the RIVE model according to a Lagrangian routing of water masses along the hydrographic network (Billen et al., 1994) and is a generic model of water quality and biogeochemical functioning of large drainage networks that simulates water quality. PyNuts is a modeling environment that can calculate the constraints (diffuse and point sources) on the Riverstrahler model at a multiregional scale (Desmit et al., 2018 for the Atlantic façade).

2.2.1. Development of an inorganic carbon module

Introducing the carbonate system

The carbonate system was described by a set of equations (named CO₂-module) based on a previous representation provided by Gypens et al. (2004) and adapted for freshwater environments (N. Gypens and A.V. Borges, personal communication). This CO₂-module was fully integrated in the RIVE model (Figure 2). It aims at computing the speciation of the carbonate system based on two new state variables: dissolved inorganic carbon (DIC) and total alkalinity (TA), making it possible to calculate carbon dioxide (CO₂). The module uses three equations (see S3: Eqs. 1, 2, 3) that also calculate bicarbonate (HCO₃⁻), carbonate (CO₃²⁻) and hydronium (H₃O⁺). Indeed, two variables of the carbonate system are sufficient to calculate all the other components (Zeebe and Wolf-Gladrow, 2001). Here, DIC and TA were selected

because the biological processes involved in their spatiotemporal variability along the aquatic continuum were already included in the RIVE model (Figure 2). We calculated pH as a function of TA and DIC using the Culberson equation (Culberson, 1980) (S3.4).

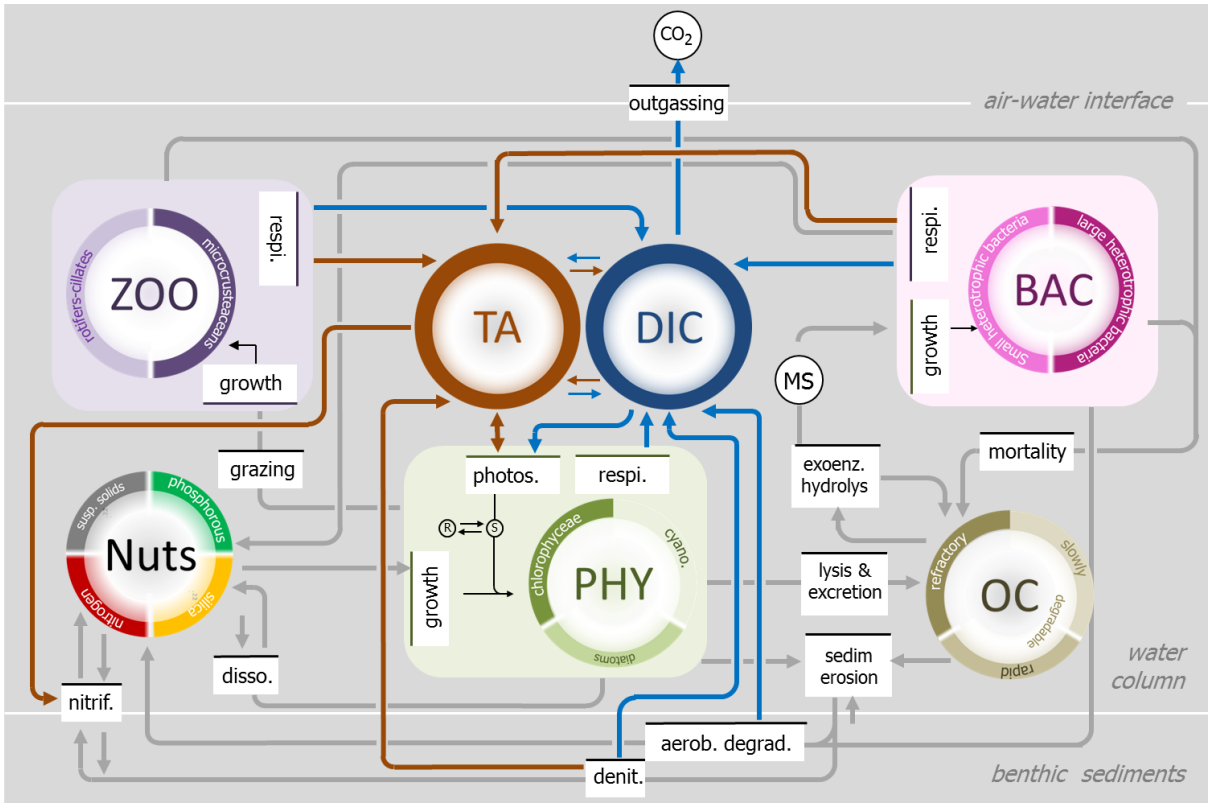


Figure 2 Schematic representation of the ecological RIVE model (inspired from Billen et al. 1994, Garnier & Billen, 1994), with grey lines indicating the main processes simulated in the water column and at the interface with sediment (oxygen not shown), and implementation of the new inorganic module, based on total alkalinity (TA, maroon) and dissolved inorganic carbon (DIC, blue).

226 Aquatic processes affecting TA and DIC

227 The exchange of CO₂ between the water surface and the atmosphere depends, respectively, on
 228 the gas transfer velocity (k-value) and on the sign of the CO₂ concentration gradient at the water
 229 surface–atmosphere interface (S3.5). Change in pCO₂ will in turn affect DIC concentrations
 230 (see Table 2, Eq. 1). Dissolved or particulate organic matter is mostly degraded by microbial
 231 activities (more or less quickly depending on their biodegradability), resulting in CO₂ and
 232 HCO₃⁻ production (Servais et al., 1995), thus inducing a change in DIC and TA concentrations
 233 in the water column (Table 2, Eq. 2, Figure 2). Photosynthesis and denitrification processes also
 234 affect DIC and TA (Table 2, Eqs. 3–5), while instream nitrification only influences TA (Table
 235 2, Eq. 6, Figure 2).

236 *Table 1 Stoichiometry of the biogeochemical processes, influencing dissolved inorganic carbon (DIC) and total*
 237 *alkalinity (TA) in freshwater, as taken into account in the new inorganic carbon module. TA and DIC expressed*
 238 *in mol:mol of the main substrate (either C or N).*

Process	Equation	DIC	TA	Eq.
FCO ₂	$CO_2(aq) \leftrightarrow CO_2(g)$	±1	0	1
Aerobic degradation	$C_{106}H_{263}O_{11}N_{16}P + 106O_2 \rightarrow 92CO_2 + 14HCO_3^- + 16NH_4^+ + HPO_4^{2-} + 92H_2O$	+1	+14/106	2
Photosynthesis (NO ₃ ⁻ uptake)	$106CO_2 + 16NO_3^- + H_2PO_4^- + 122H_2O + 17H^+ \rightarrow C_{106}H_{263}O_{11}N_{16}P + 138O_2$	-1	+17/106	3
Photosynthesis (NH ₄ ⁺ uptake)	$106CO_2 + 16NH_4^+ + H_2PO_4^- + 106H_2O \rightarrow C_{106}H_{263}O_{11}N_{16}P + 106O_2 + 15H^+$	-1	-15/106	4
Denitrification	$5CH_2O + 4NO_3^- + 4H^+ \rightarrow 5CO_2 + 2N_2 + 7H_2O$	+1	+4/5	5
Nitrification	$NH_4^+ + 2O_2 \rightarrow 2H^+ + H_2O + NO_3^-$	0	-2	6

239 **State equations and parameters of the inorganic carbon module**

240 These processes affecting TA and DIC result in equations governing inorganic carbon dynamics

241 as:

$$TA = TA_{t-1} + dt \cdot \frac{dTA}{dt} + TA_{\text{inputs}} \quad \text{Eq. 7}$$

242 with:

$$\begin{aligned} \frac{dTA}{dt} = & \left(\frac{14}{106} \frac{(respbact + respZoo + respBent)}{M(C)} \right. \\ & + \left(\frac{4}{5} Denit - 2 \cdot nitr['AOB'] \right) \cdot M(N)^{-1} + \left(\frac{17}{106} \frac{uptPhyNO_3^-}{uptPhyN} \right. \\ & \left. \left. - \frac{15}{106} \frac{uptPhyNH_4^+}{uptPhyN} \right) \cdot phot \cdot M(O_2)^{-1} \right) 1000 \end{aligned} \quad \text{Eq. 8}$$

243 where TA_{t-1} is the value of TA ($\mu\text{mol L}^{-1}$) in the previous time step (t-1). *Respbact*, *RespZoo*,
 244 and *respBent* are respectively the heterotrophic planktonic respiration of bacteria, zooplankton
 245 and benthic bacteria already included in RIVE ($\text{mgC L}^{-1} \text{h}^{-1}$). $M(C)$ is the molar mass of the
 246 carbon (12 g mol^{-1}). *Denit* and *nitr*['AOB'] are respectively the processes of denitrification and
 247 nitrification by ammonia-oxidizing bacteria (AOB) as implemented in the RIVE model (mgN
 248 $\text{L}^{-1} \text{h}^{-1}$); $M(N)$ is the molar mass of the nitrogen (14 g mol^{-1}). *phot* is the net photosynthesis
 249 ($\text{mgO}_2 \text{ L}^{-1} \text{h}^{-1}$). *uptPhyN* is the nitrogen uptake by phytoplankton ($\text{mgN L}^{-1} \text{h}^{-1}$) which is
 250 differentiated for nitrate (*uptPhyNO₃⁻*, $\text{mgC L}^{-1} \text{h}^{-1}$) and ammonium (*uptPhyNH₄⁺*, mgC L^{-1}
 251 h^{-1}), and $M(O_2)$ is the molar mass of the dioxygen (32 g mol^{-1}). TA_{inputs} is TA ($\mu\text{mol L}^{-1}$)
 252 entering the water column by diffuse sources (groundwater and subsurface discharges) and
 253 point sources (WWTPs).

$$DIC = DIC_{t-1} + dt. \frac{dDIC}{dt} + DIC_{inputs} \quad \text{Eq. 9}$$

254 with:

$$\begin{aligned} \frac{dDIC}{dt} = & (resp_{bact} + resp_{Zoo} + resp_{Bent}) + denit \cdot M(C) \cdot M(N)^{-1} \\ & + phot \cdot M(C) \cdot M(O_2)^{-1} + \frac{F_{CO_2}}{depth} \end{aligned} \quad \text{Eq. 10}$$

255 where DIC_{t-1} is the value of DIC (mgC L⁻¹) in the previous time step (t-1). F_{CO_2} is the CO₂
 256 flux at the water-atmosphere interface in mgC m⁻² h⁻¹ described in S3.5; depth is the water
 257 column depth (m).

258 The different values of constants and parameters used in the inorganic carbon module are
 259 introduced in Table 1 of S3.6. The full inorganic carbon module is described in S3 (3.1 to 3.6).

260 **2.2.2. Input constraints of the pyNuts-Riverstrahler model**

261 **Diffuse sources from soil and groundwater**

262 Diffuse sources are calculated at the scale of each modeling units, based on several spatially
 263 explicit databases describing natural and anthropogenic constraints on the Seine River basin.
 264 Diffuse sources are taken into account by assigning a yearly mean concentration of carbon and
 265 nutrients to subsurface and groundwater flow components, respectively. These concentrations
 266 are then combined with a 10-day time step description of surface and base flows to simulate the
 267 seasonal contribution of diffuse emissions to the river system. For nutrients, several applications
 268 of the Riverstrahler on the Seine River basin refined the quantification of diffuse sources: e.g.,
 269 Billen and Garnier (2000) and Billen et al. (2018) for nitrogen; Aissa-Grouz et al. (2016) for
 270 phosphorus; Billen et al. (2007), Sferratore et al. (2008) and Thieu et al. (2009) for N, P and Si.

In this study we revised our estimates for diffuse organic carbon sources and propose TA and DIC values for the Seine basin. The summary of all the carbon-related inputs of the model is provided in Table 3.

Dissolved organic carbon (DOC) input concentrations were extracted from the AESN database (<http://www.eau-seine-normandie.fr/>, last accessed 2020/02/11) and averaged by land use for subsurface sources (mean, 3.13 mgC L⁻¹; sd, 4.56 mgC L⁻¹; 3225 data for 2010–2013). For groundwater sources, concentrations were extracted from the ADES database (www.adeseaufrance.fr, last accessed 2020/02/11) and averaged by MESO waterbodies (French acronym: Masse d'Eau SOuerraine, see S4; mean, 0.91 mgC L⁻¹; sd, 0.8 mgC L⁻¹; 16,000 data for 2010–2013). These concentrations were separated into three pools of different biodegradability levels, with 7.5% rapidly, 17.5% slowly biodegradable and 75% refractory DOC for subsurface sources and 100% refractory DOC for groundwater flow (Garnier, unpublished).

Total POC inputs were calculated based on estimated total suspended solid (TSS) fluxes, associated with a soil organic carbon (SOC) content provided by the LUCAS Project (samples from agricultural soil, Tóth et al., 2013), the BioSoil Project (samples from European forest soil, Lacarce et al., 2009) and the Soil Transformations in European Catchments (SoilTrEC) Project (samples from local soil data from five different critical zone observatories (CZOs) in Europe, Menon et al., 2014) (Aksoy et al., 2016). TSS concentrations were calculated using fluxes of TSS provided by WaTEM-SEDEM (Borrelli et al., 2018) and runoffs averaged over the 1970–2000 period (SAFRAN-ISBA-MODCOU, SIM; Habets et al., 2008). The POC mean was 8.2 mgC L⁻¹; sd, 10.4 mgC L⁻¹ in subsurface runoff, and 0.8 mgC L⁻¹; sd, 1.0 mgC L⁻¹ in groundwater discharge. The same ratio of DOC reactivity was applied for three classes of POC

degradability. The kinetics for POC and DOC hydrolysis and parameters however are different (Billen and Servais, 1989; Garnier et al., 2002).

DIC and TA are brought by subsurface and groundwater discharges (Venkiteswaran et al., 2014). DIC is defined by the sum of bicarbonates (HCO_3^-), carbonates (CO_3^{2-}) and CO_2 . Unlike HCO_3^- and CO_3^{2-} measured in groundwater on a regular basis by French authorities (ADES, www.adeseaufrance.fr, last accessed 2020/02/11), CO_2 concentrations were not measured in their survey. TA values are also provided in the ADES database.

To calculate DIC concentrations in groundwaters, we therefore used our own CO_2 measurements, equaling on average $15.92 \text{ mg C L}^{-1}$; sd, 7.12 mgC L^{-1} (55 measurements in six piezometers in the Brie aquifer during 2016–2017) (see methodology in Marescaux et al., 2018a). DIC and TA were averaged for the 48 unconfined hydrogeological MESO units of the basin (see concentrations in S4) during the recent period (2010–2015), including the simulation period. In Figure 3, a summary of TA and DIC inputs by MESO units is shown by grouping MESO units according to lithology and geological ages.

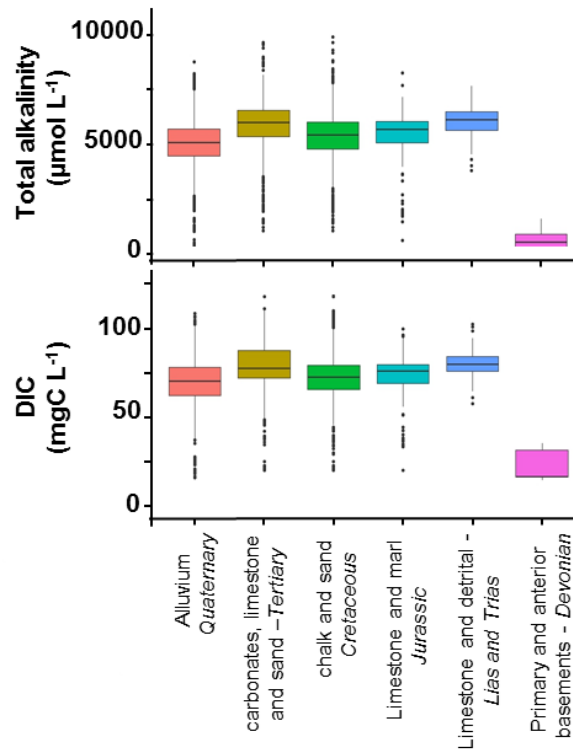


Figure 3 Boxplots of total alkalinity ($\mu\text{mol L}^{-1}$) and dissolved inorganic carbon (DIC, mgC L^{-1}) groundwater concentrations by grouping the MESO units. The lower, intermediate and upper parts of the boxes represent, respectively, the 25th, 50th and 75th percentiles and the circles represent the outlier values (source: ADES). The color code is the same as the one in S4 spatially representing the MESO units of the basin.

Documenting TA and DIC diffuse sources based on MESO units ensures a representation of their spatial heterogeneity in the Seine River basin. Carbonate waters showed higher TA and DIC mean concentrations while crystalline waters had the lowest mean concentrations in TA and DIC (primary and anterior basements from Devonian, Figure 3). Aquifers from Tertiary and alluvium from Quaternary had a more heterogeneous distribution of their concentrations (Figure 3). TA and DIC by MESO units were then spatially averaged at the scale of each modeling unit of the pyNuts-Riverstrahler model (69 modeling units, subdivided according to

Strahler ordination, S2), thus forming a semi-distributed estimate of groundwater concentrations.

TA and DIC measurements in lower order streams cannot be considered as representative of subsurface concentrations because lower order streams are expected to degas strongly in a few hundred meters, as shown for N₂O by Garnier et al. (2009) and for CO₂ in Öquist et al. (2009). We have considered similar concentrations and spatial distribution for subsurface components to those obtained for groundwater (from 25 to 92 mgC L⁻¹ DIC, and from 663 to 5580 µmol L⁻¹ TA, Figure 3).

Point sources from WWTP effluents

The pyNuts-Riverstrahler model integrates carbon and nutrient raw emissions from the local population starting from the collection of household emissions into sewage networks until their release after specific treatments in WWTPs. In the Seine River basin, most of these releases are adequately treated before being discharged to the drainage network. DOC discharge from WWTPs was described according to treatment type, ranging from 2.9 to 9.4 gC inhab⁻¹ day⁻¹ while POC discharge ranged from 0.9 to 24 gC inhab⁻¹ day⁻¹ based on the sample of water purification treatment observed in the Seine basin (Garnier et al., 2006; Servais et al., 1999).

TA and DIC were measured at eight WWTPs selected to reflect various treatment capacities (from 6 10³ inhab eq to 6 10⁶ inhab eq) and different treatment types (activated, sludge, Biostyr® Biological Aerated Filter) in the Seine River basin. Sampling and analysis protocols are provided in S5. This sampling did not allow us to highlight differences in per capita TA and DIC emissions. Consequently, we used a fixed value of 3993 µmol L⁻¹ for TA and 70 mgC L⁻¹ for DIC, which correspond to the weighted mean by WWTP capacity of our measurements and are in agreement with values from Alshboul et al. (2016) found in the literature.

Impact of the reservoirs

Nutrients and organic carbon cycling within the three reservoirs of the Seine River network were simulated using the biogeochemical RIVE model adapted for stagnant aquatic systems (Garnier et al., 1999). Owing to the absence of an inorganic carbon module in the modeling of reservoirs yet, we used mean measurements of TA and DIC in reservoirs as forcing variables to the river network. The Der lake reservoir was sampled three times (2016/05/24, 2016/09/12, 2017/03/16) and among others, TA and DIC were measured (see Table 3). Recent sampling campaigns showed that TA and DIC are similar for the three reservoirs (X. Yan, pers. comm.).

Table 2 Summary of the carbon related inputs of the pyNuts-Riverstrahler model.

Input variables	Flow	Database	averaged	values	source
DOC	subsurface	AESN	land use	mean: 3.13 mgC L ⁻¹ ; sd: 4.56 mgC L ⁻¹ ;	http://www.eau-seine-normandie.fr/
	groundwater	ADES	MESO units	mean: 0.91 mgC L ⁻¹ ; sd: 0.8 mgC L ⁻¹	www.adeseaufrance.fr
POC	subsurface	LUCAS, BioSoil and SoilTrEC Projects	based on estimated total suspended solids (TSS) fluxes, associated with a soil organic carbon (SOC) content	mean: 8.2 mgC L ⁻¹ , sd: 10.4 mgC L ⁻¹	(Aksoy et al., 2016)
	groundwater			mean: 0.8 mgC L ⁻¹ , sd: 1.0 mgC L ⁻¹	
DIC	subsurface	ADES	MESO units	from 25 to 92 mgC L ⁻¹	www.adeseaufrance.fr
	groundwater			from 25 to 92 mgC L ⁻¹	
TA	subsurface	ADES	MESO units	from 663 to 5580 μmol L ⁻¹	www.adeseaufrance.fr
	groundwater			from 663 to 5580 μmol L ⁻¹	
DOC	Point sources	Measurements	According to WWTP treatment and capacity	2.9 to 9.4 gC inhab ⁻¹ day ⁻¹	(Garnier et al. 2006; Servais et al. 1999)
POC	Point sources	Measurements		0.9 to 24 gC inhab ⁻¹ day ⁻¹	
DIC	Point sources	Measurements	weighted mean by WWTP capacity	70 mgC L ⁻¹	This study
TA	Point sources	Measurements	weighted mean by WWTP capacity	3993 μmol L ⁻¹	This study
DIC	Reservoirs	Measurements in the Der Lake	by year	mean: 23 mgC L ⁻¹ , sd: 4 mgC L ⁻¹	This study
TA	Reservoirs	Measurements in the Der Lake	by year	mean: 1890 μmol L ⁻¹ , sd: 350 μmol L ⁻¹	This study

2.2.3. Observational data

We selected the 2010–2013 timeframe for setting up and validating the new inorganic module. This period includes the year 2011, which was particularly dry in summer (mean annual water discharge at Poses, 366 m³ s⁻¹) and 2013, which was wet (mean annual average water discharge at Poses, 717 m³ s⁻¹) while 2010 and 2012 showed intermediate hydrological conditions (mean annual average water discharges at Poses, 418 m³ s⁻¹ and 458, m³ s⁻¹, respectively) (data source: Banque Hydro).

The pCO₂ values (ppmv) were calculated using CO2SYS software algorithms (version 25b06, Pierrot et al., 2006) based on existing data collected by the AESN. TA, pH, and water

temperature data sets were used for the 2010–2013 selected period (8693 records for these three variables, i.e., around 1209 stations distributed throughout the Seine basin, measurements that were taken at a fixed time – 9:00-15:00 UTC–, and could not represent diurnal fluctuations). The carbonate dissociation constants (K1 and K2) applied were calculated from Millero (1979) with zero salinity and depending on the water temperature. Because pCO₂ calculations from pH and TA can lead to overestimation of pCO₂ (Abril et al., 2015), the pCO₂ calculated data were corrected by a relationship established for the Seine River and based on pCO₂ field measurements (Marescaux et al., 2018b). To compute the interannual average over the 2010–2013 period, data were averaged monthly, then annually at each measurement station and then spatially averaged (i.e., by Strahler orders). Four stations offering sufficient data for the 2010–2013 period were selected for appraising seasonal patterns. They are located along the main stem of the Marne-Lower Seine River: Poses (the outlet), Poissy (downstream of the SAV WWTP), Paris and Ferté-sous-Jouarre (upstream of Paris) (Figure 1a).

All data were processed using R (R Core team, 2015) and QGIS (QGIS Development Team, 2016). Kruskal-Wallis tests were used to compare simulated and measured pCO₂ averages.

2.2.4. Evaluation of the model

Root mean square errors normalized to the range of the observed data (NRMSE) were used to evaluate the pyNuts-Riverstrahler model including the inorganic module, indicating the variability of the model results with respect to the observations, normalized to the variability of the observations. NRMSE analysis were performed on inter-annual variations per decade for the 2010-2013 period, combining observations and simulations at four main monitoring stations along the longitudinal profile of the Seine River: Poses, Poissy (downstream of Paris), Paris, and Ferté-sous-Jouarre (upstream of Paris).

3. Results

3.1. Simulations of spatial and seasonal variations of $p\text{CO}_2$.

3.1.1. CO_2 from lower order streams to larger sections of the Seine River

Simulations of CO_2 concentrations averaged for 2010–2013 by Strahler orders showed that pyNuts-Riverstrahler succeeded in reproducing the general trends of CO_2 observations (7565 data) (Figure 4). Although differences in CO_2 concentrations between the different order streams were not significant, their means tended to decrease from lower order streams (SO1) (width < 100 m) to SO5, and to finally increase in the higher order streams (width > 100 m) from SO6 to SO7, downstream of the Paris conurbation. Some discrepancy appeared for order 1, with simulations yielding higher values than the observations while for orders 2–7 simulation values were conversely lower than observation values. The corresponding k -values calculated for the Seine ranged from 0.04 to 0.23 m h^{-1} with higher values in the first streams and lower values in larger rivers (not shown), with CO_2 outgassing positively related to the k -value (S3.5 Eq. S25).

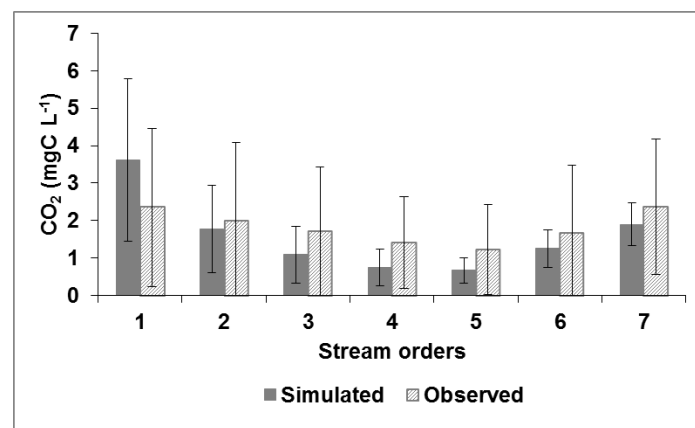


Figure 4 Carbon dioxide concentrations in the Seine waters (CO_2 , mgC L^{-1}) simulated by the pyNuts-Riverstrahler model (dark gray) and observed (light grey) as a function of the stream order averaged over the 2010–2013 period (whiskers indicating standard deviations).

3.1.2. Profiles of the main stem Marne and Lower Seine (at Poses)

In the same period (2010-2013), a focus on the main stem from the Marne River (SO6) until the outlet of the Seine River (Poses, SO7) showed that the model correctly reproduced longitudinal variations. Higher concentrations of CO₂ downstream of Paris, and a peak of CO₂ concentrations immediately downstream of the SAV WWTP were followed by a progressive decrease until the estuary (Figure 5). Note that the estuarine CO₂ concentrations were specifically modeled by Laruelle et al. (2019), using these outputs of the Riverstrahler simulations.

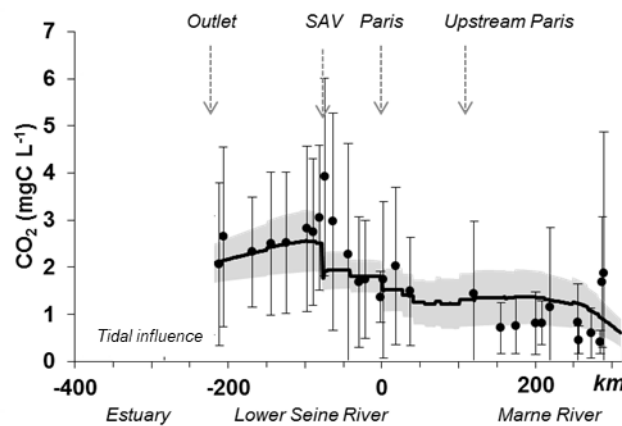


Figure 5 Observed (dots) and simulated (line) mean carbon dioxide concentrations (CO₂, mgC L⁻¹) along the main stem of the Marne River (km -350 to 0) and the lower Seine River (km 0–350) averaged over the 2010–2013 period. The simulation envelope (gray area) represents standard deviations of simulated CO₂ concentrations. Whiskers are standard deviations between observed CO₂ concentrations.

3.1.3. Seasonal variations

Upstream, within Paris, and downstream of Paris, the model provides simulations in the right order of magnitude of the observed CO₂, DIC, TA and pH values, despite the fact that TA was underestimated in the two upstream stations selected for all seasons (Figure 6). DIC and TA

422 simulations followed the observed seasonal patterns with a depletion of concentrations
423 occurring in summer/autumn related to low-flow support by the reservoirs. Indeed, reservoirs
424 showed lower TA and DIC concentrations than rivers (Table 3). In addition to the intra-/inter-
425 stream order variabilities of CO₂ (Figure 4), CO₂ concentrations showed a wide spread in values
426 over the year (Figure 6). Although simulated CO₂ concentrations fitted rather well with the
427 level of the observations (NRMSE = 15%), the model tended to overestimate the winter values
428 upstream and within Paris (Figure 6, left).

429 For DIC, simulations upstream from Paris (Figure 6, right) seemed lower than the observations
430 (but summer data are missing); however, downstream at the other three stations selected,
431 simulations accurately represented the observations (Figure 6, NRMSE = 15%). Seasonal
432 variations of TA were satisfactorily reproduced by the simulations, although they were slightly
433 underestimated by the model at the stations upstream and downstream of Paris (Figure 6,
434 NRMSE = 25%). Regarding pH, simulations were in a similar range as the observations (range,
435 7.5–8.5), and lower summer pH values in the lower Seine were correctly simulated by the model
436 (Figure 6, NRMSE = 17%).

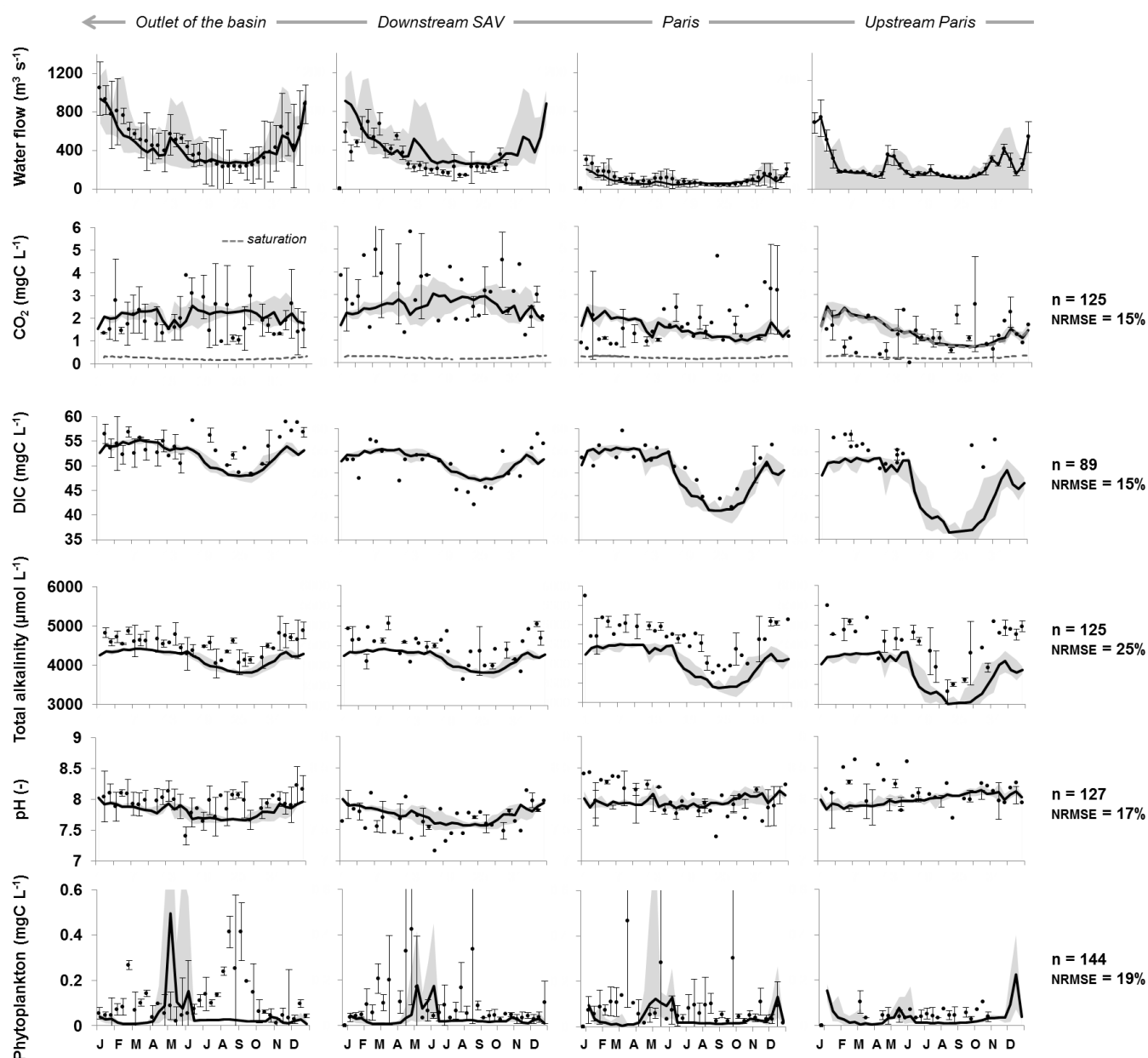


Figure 6 Ten-day simulated (lines) and observed (dots) water discharges over the 2010–2013 period (Q , $\text{m}^3 \text{s}^{-1}$), concentrations of carbon dioxide (CO_2 , mgC L^{-1} , and $\text{CO}_2 \text{ sat}$, mgC L^{-1}), dissolved inorganic carbon (DIC , mgC L^{-1}), total alkalinity (TA , $\mu\text{mol L}^{-1}$), pH (-), and phytoplankton (mgC L^{-1}). Four monitoring stations of interest along the main stem of Marne-lower Seine are shown: Ferté-sous-Jouarre (upstream of Paris on the Marne River), Paris on the lower Seine (upstream at Charenton), downstream of the SAV WWTP, and at the outlet of the basin (Poses). NRMSE analysis were performed on inter-annual variations per decade for the 2010-2013 period, combining observations and simulations at four main monitoring stations. Simulation envelope corresponds to standard deviations (gray area). For observed data, whiskers are standard deviations.

Although the level of phytoplankton biomass was adequately simulated, the summer bloom observed at the outlet was not reproduced, whereas the early spring bloom observed in the lower Seine was simulated with a time lag compared to the observations (Figure 6, bottom, NRMSE = 19%).

3.1.4. Selection of a gas transfer velocity

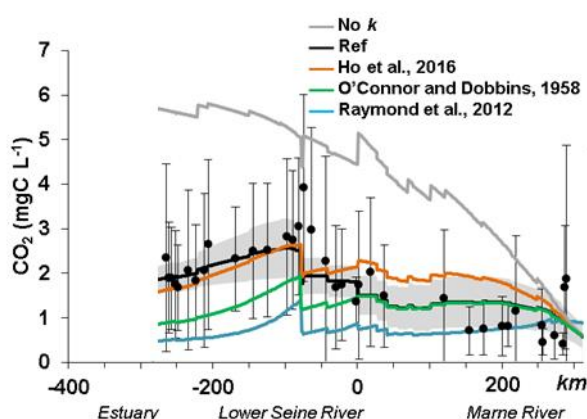


Figure 7 Influence of the gas transfer velocity formalisms along the main stem of the Seine River basin (Marne – Lower Seine River) impacted riverine CO₂ concentrations.

The way of taking into account the gas transfer velocity in the modeling approach could explain these discrepancies in SO6 and SO7 (Figure 4). Different values of k were explored specifically in the downstream part of the Seine river network (SO6 and SO7 where river width exceeds 100m) (Figure 7). Indeed, the gas transfer velocity value reported by Alin et al. (2011) was used for streams and rivers up to 100 m wide, as they recommended. Whereas these k -values provided adequate simulations in the river up to 100 m wide, for river widths greater than 100 m, we tested different k -values. In larger stream orders, we showed that calculations of k according to the Equation 5 of Table 2 by Raymond et al. (2012), induced a too high outgassing while when not using any k -value for these larger rivers, the opposite behavior with a much too low outgassing of CO₂ was observed.

Therefore, for river widths greater than 100 m, a k_{600} equation based on O'Connor and Dobbins, (1958) and Ho et al. (2016), neglecting the term related to the wind, and providing the most accurate CO₂ concentrations, was selected (see S3 for more information's on the selection of k and the tests performed):

Although these results can be improved, organic and inorganic carbon and total alkalinity budgets can be calculated at the scale of a whole drainage basin for the first time.

3.2. Alkalinity, inorganic and organic carbon budgets

We established an average inorganic and organic budget for the period studied (2010–2013) (Table 4). The budget of inorganic and organic carbon (IC and OC) of the entire Seine River basin (from headwater streams to the beginning of the estuary) showed the high contribution of external inputs (sum of point and diffuse sources accounted for 92% and 68% of IC and OC inputs, respectively) and riverine exports (68% and 66% of IC and OC outputs, respectively). These exports were at least one order of magnitude higher for the IC budget (Table 4). The substantial contribution of the Seine aquifer water flow led the IC flux brought by groundwater to dominate over those from the subsurface (respectively, 57.5% vs. 34% of total IC inputs, respectively), while for OC, the subsurface contributions were higher than the groundwater contributions (54% vs. 14% of the total OC fluxes).

Interestingly, the relative contributions of point sources to OC inputs were higher than for IC (23% and 7% of the OC and IC inputs, respectively) (Table 4).

Heterotrophic respiration by microorganisms accounted for only 1.5% of the IC inputs. Similarly, IC losses by net primary production also accounted for a small proportion, i.e., 0.6% of the IC inputs. For the OC budget, despite a contribution of autochthonous inputs from instream biological metabolisms (NPP and nitrification, 9% of inputs, and heterotrophic

487 respiration, 7%), which was relatively high compared with their proportion in IC fluxes (2.3%),
488 allochthonous terrestrial inputs still dominated the OC budget (Table 4).

489 The Seine River, at the outlet, exported 68% of the IC entering or produced in the drainage
490 network, and 66% of the OC brought to the river (including both particulate and dissolved
491 forms) (Table 4). Instream OC losses were related to heterotrophic respiration (7%) and to a
492 net transfer to the benthic sediment compartment, including sedimentation and erosion
493 processes (estimated at 28% of losses). In the IC budget, CO₂ emissions were a substantial
494 physical process (31% of the overall losses) (Table 4).

495 A similar calculation was performed for the total alkalinity (TA) budget. As for inorganic
496 carbon, the contribution of internal processes remained relatively low compared with the high
497 levels of TA in lateral inputs (diffuse sources: 93 %; point sources: 6 %) and flows exported to
498 the basin outlet (97 %). Indeed, instream production mostly relied on heterotrophic respiration
499 (< 1%) while denitrification was negligible. Photosynthesis might also produce or consume
500 alkalinity whether NO₃⁻ or NH₄⁺ is the preferential N source of phytoplankton's uptake, but in
501 our budget it resulted in our budget in a net TA reduction (2%), while nitrification also
502 contributed to less than 1% of TA output.

*Table 4 Budget of the Seine hydrosystem for inorganic and organic carbon (kgC km⁻² yr⁻¹) and total alkalinity (TA, mol km⁻² yr⁻¹) as calculated by the pyNuts-Riverstrahler model averaged over the period 2010-2013. * TA input related to NPP refers to the net difference between TA produced by photosynthesis on NO₃ uptake and photosynthesis on NH₄ uptake (reducing alkalinity). **Net sediment loss is the difference between the erosion and the sedimentation calculated by the model.*

2010-2013	Processes involved in inorg C budget	kgC km ⁻² yr ⁻¹	%
Input to river	Diffuse sources from subroot	5963	34.4
	Diffuse sources from groundwater	9968	57.5
	Urban point sources	1135	6.6
	Heterotrophic respiration	266	1.5
	Denitrification	0	0.0
Output from river	Delivery to the outlet	12483	68.4
	CO ₂ emissions	5619	30.8
	Nitrification	37	0.2
	NPP	105	0.6
2010-2013	Processes involved in TA budget	mol km ⁻² yr ⁻¹	%
Input to river	Diffuse sources from subroot	360983	34.9
	Diffuse sources from groundwater	604145	58.4
	Urban point sources	66770	6.4
	Heterotrophic respiration	2972	0.3
	Denitrification	0	0.0
Output from river	Delivery to outlet	1004299	97.1
	Nitrification	6219	0.6
	NPP *	24352	2.4
2010-2013	Processes involved in org C budget	kgC km ⁻² yr ⁻¹	%
Input to river	Diffuse sources from subroot	870	53.9
	Diffuse sources from groundwater	227	14.1
	Urban point sources	375	23.2
	Nitrification	37	2.3
	NPP	105	6.5
Output from river	Delivery to the outlet	1086	65.7
	Heterotrophic respiration	110	6.7
	Net sedimentation **	456	27.6

3.3. Carbon aquatic processes

Whereas IC and OC budgets of the Seine hydrosystem were clearly dominated by external terrestrial inputs and outputs through deliveries at the coast, an attempt was made here to analyze instream processes involved in the IC and OC cycles (Figure 8, Figure 9).

The average spatial distribution of IC processes, as calculated by the model, was mapped for the 2010–2013 period (Figure 8). Benthic activities were the greatest in smaller streams. By contrast, net primary production and heterotrophic planktonic respiration, which both followed a similar spatial pattern, increased as Strahler order increased, reaching their highest values in the lower Seine River. All these biological processes involved in the IC cycle were therefore highly active in the main stem of the river, while on the other hand CO₂ outgassing occurred mainly in the basin's small headwater streams (Figure 8).

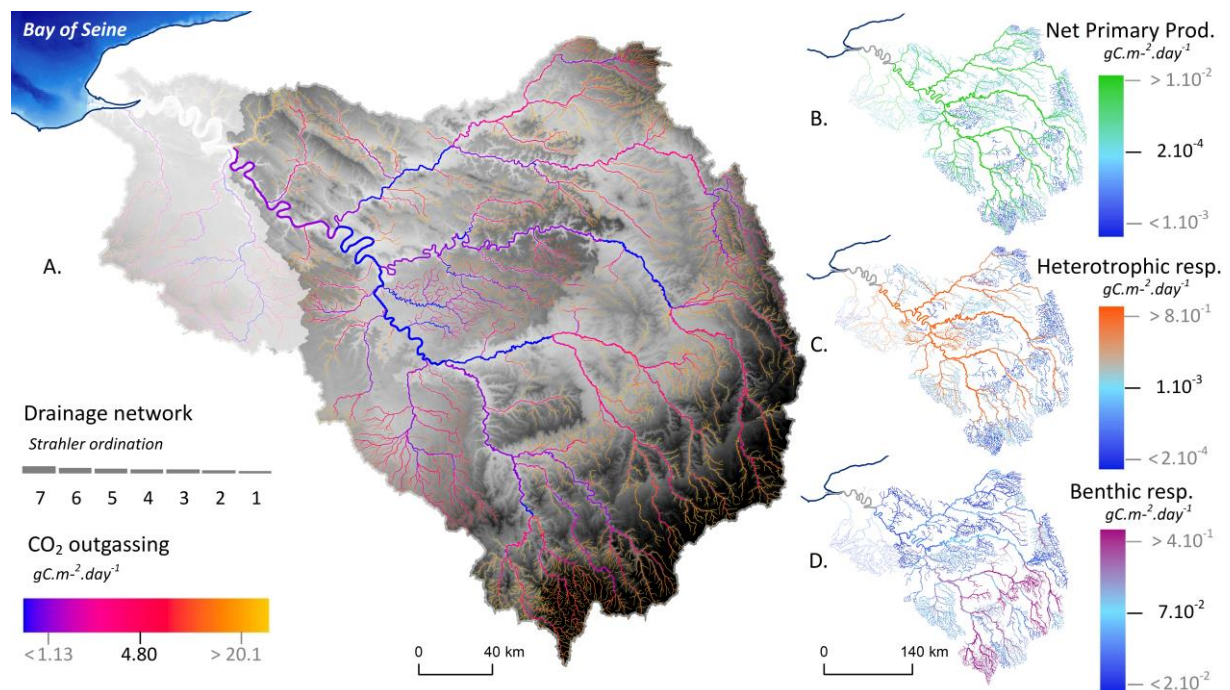


Figure 8 Instream processes involved in the inorganic carbon cycle simulated by pyNuts-Riverstrahler and averaged over the 2010–2013 period for the Seine River network until its fluvial outlet at Poses. a) CO₂ outgassing (blue–yellow, gC m⁻² day⁻¹); b) net primary

production (blue–green, gC m⁻² day⁻¹); c) heterotrophic planktonic (blue–violet); d) benthic respiration (blue–orange, gC m⁻² day⁻¹) are represented in the hydrographic network.

Regarding the OC processes, mostly linked to biological activity, they were analyzed in terms of ecosystem metabolism (Figure 9). The net ecosystem production (NEP, gC m⁻² day⁻¹) is defined as:

$$\text{NEP} = \text{NPP} - \text{Het. Respiration}$$

where NPP is the net primary production (gC m⁻² day⁻¹) depending on the growth of phytoplankton. NPP contributes to building phytoplankton biomass that constitutes a stock of organic carbon, emitted in turn as CO₂ by respiration (Het. respiration, gC m⁻² day⁻¹).

Simulations showed that NEP would remain negative in the entire drainage network (Figure 9). However, NEP must be analyzed with caution since the phytoplankton pattern was not adequately represented (see Figure 6). In SO1, this negative NEP was associated with almost no NPP, and heterotrophic respiration was dominated by benthic activities (see Figure 8). In SO5, NEP was less negative than in SO1 (Figure 9), and heterotrophic respiration was lower than in SO1 while NPP was higher. In the lower Seine River (SO7), NPP increased as did heterotrophic respiration, which reached its highest value in this downstream stretch receiving treated effluents from WWTPs. Therefore, the increase in NPP did not result in positive NEP. The entire drainage network was thus supersaturated in CO₂ with respect to atmospheric concentrations, and constituted a source of CO₂. This supersaturation was the highest in smaller orders, lower in intermediate orders and increased again in the lower Seine River (Figure 4, see also Figure 8).

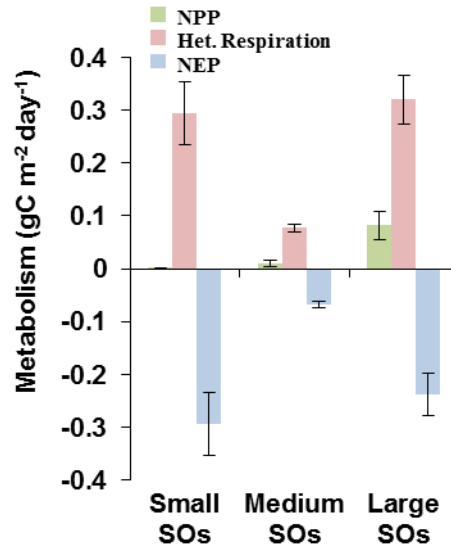


Figure 9 Metabolism for small, intermediate and large stream orders (SO) (here represented by SO1, SO5, and SO7, respectively) of the Seine basin simulated by pyNuts-Riverstrahler and averaged over the 2010–2013 period. Net primary production (NPP, gC m² day⁻¹), heterotrophic respiration (Het. respiration, gC m² day⁻¹), net ecosystem production (NEP, gC m² day⁻¹).

4. Discussion

4.1. Evaluation of the model

Simulated CO₂ concentrations tend to be higher than observed ones for SO1. These differences may be related to the high variability of CO₂ in SO1, and the scarcity of measurements in spring. However, Öquist et al. (2009) estimated that up to 90% of daily soil DIC import into streams was emitted to the atmosphere within 200 m. Such a CO₂ emission pattern can be applied to the Seine, as a similar result was found for N₂O (Garnier et al., 2009). Since soil emissions were very difficult to capture, we considered that concentrations in groundwater (DIC and TA) closely reflect the composition of diffuse sources, much like soil composition. This assumption probably underestimates the DIC/TA ratio brought to the river in lower order streams.

Differently from SO1, simulated concentrations in SO2–7 are lower than the observed values (Figure 4). Overall, the NRMSE indicating a percentage of variation was less than 20%, except for TA (25%).

Regarding gas transfer velocity values, an equation for large rivers with no tidal influence using wind speed could be more appropriate (Alin et al., 2011) and could decrease NRMSE in these downstream sections of the river. However, the Riverstrahler model does not consider wind as an input variable, which would have required the model to have a much higher spatiotemporal resolution to reflect its spatiotemporal heterogeneity in the Seine basin, with for example, the diurnal cycle affected by phenomena such as breezes (Quintana-Seguí et al., 2008).

Future work with direct k measurements and/or a new representation of k -values in the model could help improve outgassing simulations with pyNuts-Riverstrahler. A test of different k formulations on high stream orders (width > 100 m) representing only 1.5% of the length of the river system showed an increase of the total CO₂ outgassing estimates by up to 6.2%. Our model is k sensitive and our estimates differs from the results of Lauerwald et al. (2017), who observed that a large variation in k does not lead to a significant change in simulated aquatic CO₂ emissions. For the Seine River here, we indeed used a more accurate k -value calculated at each time step (10 days) and at every kilometers of the river network (according to water temperature, velocity, depth). In addition, a huge organic carbon load is brought by WWTPs in this Seine urbanized hydrosystem that disrupts carbon dynamics (e.g., WWTPs treating 12 million inhab. eq in the Parisian conurbation) in the downstream part of the Seine River, in contrast to simulations on a natural network (Lauerwald et al., 2017).

Regarding seasonal patterns, DIC and alkalinity amplitudes were suitably captured and the level of the values was correct. DIC and TA observations showed a strong decrease from June/July to November (maximum amplitude decrease, 10 mgC L^{-1} and $1000 \text{ } \mu\text{mol L}^{-1}$), as illustrated by the model. For the Seine River, the water flow decrease in summer was mainly related to the decrease in runoff water, meaning that the groundwater contribution was comparatively higher at this time. According to our measurements, these groundwaters were more concentrated in TA, DIC, and CO_2 than runoff water. However, water released by upstream reservoirs (supporting low flow in the downstream section of the Seine network) account for a significant proportion of the river discharge during summer and was characterized by lower TA, DIC and CO_2 concentrations. Then the decrease observed was related to the contribution of reservoirs. These results strongly encourage the implementation of an inorganic carbon module in the modeling of reservoirs, already coupled with Riverstrahler for nutrients and organic carbon (Garnier et al., 1999).

The model showed a weak performance in representing CO_2 seasonality. Referring to a previous study (Marescaux et al., 2018b), pCO_2 seasonality in the Seine River resulted from a combination of water temperature and hydrology leading to an increase in pCO_2 and CO_2 evasion fluxes from winter to summer/autumn. The pyNuts-Riverstrahler model however has an accurate representation of these constraints and would not account for these discrepancies. Also, despite the fact that the biomass level of phytoplankton was consistent with the observations, the seasonal pattern was not satisfactory reproduced by the model. However, it is worth mentioning that phytoplankton parameters in RIVE were determined through laboratory experiments at a time when the amplitude of algal blooms was much higher than at present (up to $4.5\text{-}6 \text{ mgC L}^{-1}$ i.e., chlorophyll *a* reaching $150 \text{ } \mu\text{gChla L}^{-1}$, Garnier et al., 1995). Indeed, the

implementation of the European Water Framework Directive in the 2000s with enhancement of treatments in WWTPs greatly improved water quality (Romero et al., 2016). New laboratory experiments for possibly taking into account additional phytoplankton groups or species in these new trophic conditions and/or mixing stochastic and mechanistic modeling are required to better represent phytoplankton temporal dynamics in the model. In addition, the observed incident light, instead of the empirical relationship used, would improve the early winter bloom, newly occurring in a changing environment.

4.2. Export fluxes

The new implementation of an inorganic carbon module in the pyNuts-Riverstrahler model allows us to estimate CO₂ outgassing of the Seine River at $364 \pm 99 \text{ GgC yr}^{-1}$ ($1.4 \text{ GgC km}^{-2} \text{ yr}^{-1}$ taking into account a river surface area of 260 km^2). This is significantly lower than in our previous estimate of 590 GgC yr^{-1} ($2.2 \text{ GgC km}^{-2} \text{ yr}^{-1}$ from a river surface area of 265 km^2) using CO₂ measurements only (Marescaux et al., 2018a). This difference is explained by various factors. Marescaux et al (2018a) use k formulates according to Raymond et al. (2012, Eq. 5 in Table 2) all along the Seine drainage network and consequently, the value of CO₂ emissions was most likely overestimated (see 4.1. Evaluation of the model). We also acknowledged that the CO₂ outgassing estimate yielded by simulations might overall slightly underestimate emissions with respect to Figure 4, which showed that our simulated CO₂ concentrations were overestimated for SO1 but underestimated for SO2 to SO7. In the model, a better spatio-temporal resolution and description of the water temperature, the water velocity and a more accurate description of the k -value adopted here with different k -values for small and high stream orders would be associated with less outgassing than in our previous study. For

this reasons, we believe that our estimate of 364 ± 99 GgC/yr, using our process based model is a more accurate value of CO₂ emissions from the Seine River.

The outgassing found for the Seine River by surface area of river of 1400 ± 381 gC m⁻² yr⁻¹ is in the middle range of the average estimates of outgassing from temperate rivers (70-2370 gC m⁻² yr⁻¹), including the St. Lawrence River (Yang et al., 1996), Ottawa River (Telmer and Veizer, 1999), Hudson River (Raymond et al., 1997), US temperate rivers (Butman and Raymond, 2011) and Mississippi River (Dubois et al., 2010). This high variability for these temperate rivers is strongly dependent on whether or not the first-order streams were considered in the outgassing. Similar to our study, Butman and Raymond (2011) took into account lower order streams and rivers while lower estimates correspond to studies investigating large rivers, excluding lower order streams. Indeed, outgassing are often greater in headwater streams than in large rivers owing to higher CO₂ concentrations and headwater streams have higher gas transfer velocities (Marx et al., 2017; Raymond et al., 2012a). The mapping of CO₂ outgassing in the Seine basin clearly showed these spatial trends, with smaller streams releasing more CO₂ than median and larger rivers (see Figure 8). Indeed, first-order streams of the Seine River represents 9.6% of the Seine surface area and contributed to 40% of the total CO₂ emissions by the river network.

Regarding organic carbon, Meybeck (1993) estimated the DOC export to the ocean for a temperate climate at 1.5 gC m⁻² yr⁻¹, a value that is higher than our OC estimate of 1.1 gC m⁻² yr⁻¹ for the Seine River basin, before entering the estuarine section. Compared with other temperate rivers, the rivers of the northern France, and specifically the Seine River here, are rather flat, their low altitude limiting erosion (Guerrini et al., 1998). In addition, since the implementation of the European Water Framework Directive in the 2000s, decreasing nutrients and carbon in wastewater effluents discharged into the rivers (Rocher and Azimi, 2017),

together with a decrease in phytoplankton biomass development (Aissa Grouz et al., 2016; Romero et al., 2016) can explain this difference in DOC fluxes for the Seine, a change probably valid for many other western European rivers (Romero et al., 2013). Furthermore, the CO_2/OC ratio of the export to the estuary of the Seine hydrosystem is 5.2, which is higher than this ratio for the Mississippi River, for example (4.1; Dubois et al., 2010b; Li et al., 2013) and may be related to considerable outgassing from headwater streams taken into account in our study. Note, however, that the small Seine River basin exports only $70 \pm 99 \text{ GgC yr}^{-1} \text{ OC}$ compared with the large Mississippi River with exports amounting to $2435 \text{ GgC yr}^{-1} \text{ OC}$ (Dubois et al., 2010), and with a surface area more than 40 times greater than the Seine. Interestingly, the Seine River export was estimated at three times less than the export calculated in 1979 (250 GgC yr^{-1} , Kempe, 1984). This difference in DOC concentrations in the Seine River would be 2.8 times lower than in the 1990s (Rocher and Azimi, 2017).

We estimated the DIC export of the Seine River at $820 \pm 220 \text{ GgC yr}^{-1}$, a value higher than basins of the same size or even larger (e.g., Ottawa River, drainage area, $149,000 \text{ km}^2$, 520 GgC yr^{-1} , Telmer and Veizer, (1999); Li et al. (2013)). The high concentrations of HCO_3^- in the Seine basin already documented and related to the lithology of the Seine basin (limestone and gypsum beds from Cretaceous and Tertiary) (Kempe, 1982; 1984) may explain this high export to the river outlet. With both high CO_2 and DIC exports, the ratio of CO_2/DIC exports from the Seine River is the same as the overall ratio here (0.5, Li et al., 2013)..

4.3. Metabolism

Model simulations with the new inorganic carbon module can be used to analyze spatial variations of CO_2 in regard to instream metabolism activities. We observe that the influence of the metabolism activities on the CO_2 outgassing is low. Indeed, in the carbonated Seine River,

the IC originating from groundwater supports the CO₂ outgassing along the network (Figure 8). Nevertheless, instream metabolism activities produce or consume CO₂.

The model highlights the importance of benthic activities in headwater streams (Figure 8) that decreased downstream as heterotrophic planktonic activities increased in larger rivers, a typical pattern described by the river continuum concept (RCC, Vannote et al., 1980) and quantified for the Seine River (Billen et al., 1994; Garnier et al., 1995; Garnier and Billen, 2007). These results are also in agreement with those reported by Hotchkiss et al. (2015), who suggested that the percentage of CO₂ emissions from metabolism increases with stream size while CO₂ emissions of lower-order streams are related to allochthonous terrestrial CO₂. Regarding headwater streams, Battin et al. (2009b) described benthic activities as the highest (as also observed in our study, Figure 8) where microbial biomass is associated with streambeds characterized by exchanges with subsurface flow bringing nutrients and oxygen and increasing mineralization.

Mean NEP would remain negative in the entire basin resulting from heterotrophic conditions producing CO₂ (Figure 8 and Figure 9). However, even though the level of phytoplankton biomass was correctly simulated, the summer downstream bloom, which was not reproduced by the model, could lead to some NPP underestimation. As expected, NPP in lower order streams was lower than in higher SOs owing to shorter water residence times. Benthic respiration of lower order streams was significant (Figure 8) and made NEP highly negative. Also, small SOs were the most concentrated in CO₂ owing to the groundwater contribution. Intermediate stream orders showed the smallest CO₂ or heterotrophic respirations with NEP less than $-0.1 \text{ gC m}^{-2} \text{ day}^{-1}$. This can be explained by an increase of NPP due to a lower dilution rate than the phytoplankton growth rate (Garnier et al., 1995), and to a reduced ratio of the bottom sediment-to-water column volume, decreasing heterotrophic respiration. In higher

stream orders both NPP and heterotrophic respiration were the highest, however, they led to negative NEP lower than SO1 (Figure 8 and Figure 9). Despite photosynthesis reducing the CO₂ concentrations (Figure 6), the highest SOs were affected by wastewater effluents, resulting in an overall negative NEP.

During the recent 2010–2013 period studied herein, and in all SOs, the NPP never exceeded heterotrophic respiration (ratio NPP:Het.-Resp or P:R < 1) (Figure 9). Whereas in the past the eutrophication of the Seine River led to a P:R ratio greater than 1 in large rivers, at least during spring blooms, with P and R values increasing up to 2.5 gC m⁻² day⁻¹ (Garnier and Billen, 2007), the P:R ratio is now systematically lesser than 1. These changes, linked to an overall decrease in biological metabolism, are explained by improvements of treatments in WWTPs decreasing the organic carbon load discharged into rivers and the associated pollution, and hence decreasing the CO₂ concentration along the main stem of the Seine River (Marescaux et al., 2018b). Beside DOC, improvements wastewater treatments also reduced nutrient inputs to the river, especially phosphates, today a limiting nutrient to algal development in SO5 and 6, reducing algal peaks by a factor of 3.

5. Conclusion

The pyNuts-Riverstrahler model of biogeochemical river functioning newly includes the processes involved in the inorganic carbon cycle in order to represent the spatial dynamics and seasonal variations of CO₂ concentrations and outgassing along the Seine hydrosystem. The sensitivity of simulations to different gas transfer velocity values highlighted the need for additional refinement for the Seine River so as to choose the best model equation. In addition, revisiting the phytoplankton description in the model could facilitate a better simulation of the temporal dynamics of phytoplankton. Further, an explicit representation of the anaerobic

723 reduction chain of the benthos could enable us to specify the benthic impact on TA and DIC in
724 a greater variety of ecosystems.

725 CO₂ concentrations appear to be controlled differently along the Seine hydrosystem. In small
726 orders, concentrations were mainly driven by diffuse sources. In larger rivers, in addition to the
727 influence of groundwater and low-flow support by upstream reservoirs, concentrations showed
728 patterns linked to hydrosystem metabolisms. Indeed, blooms tended to decrease CO₂
729 concentrations, although the hydrosystem remained heterotrophic and supersaturated with
730 respect to the atmospheric CO₂ concentrations. Heterotrophic respiration increased CO₂
731 concentrations with peaks downstream of WWTP effluents enriched in organic carbon.

732 Our Riverstrahler modeling has shown that there are many factors that control CO₂ emissions
733 in basins affected by human activity along an aquatic continuum. Once validated by field
734 measurements, which are still too scarce, this generic modeling approach can be applied to any
735 drainage system to better quantify lateral CO₂ emission on a continental scale.

Data availability

The datasets generated during the current study are available from the corresponding author on reasonable request.

Author contribution

All the authors contributed to the design of the study. J.G. and V.T. are co-supervisors of the PhD. A.M. participated as a PhD student in the field campaigns, lab chemical analyzes and implementation of the new inorganic carbon module. N.G. and M.S. provided technical and scientific support for the modelling. A.M. wrote the first draft of the manuscript, and all the co-authors helped to interpret the data and write the article.

Competing interests statement

The authors declare no competing financial or non-financial interest.

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759 Josette Garnier (Research Director at the Centre National de la Recherche Scientifique, France)
760 are co-supervisors of the PhD. Nathalie Gypens is Professor at the Université Libre de Bruxelles
761 (Belgium). Marie Silvestre is GIS Engineer at the Centre National de la Recherche Scientifique
762 (France).

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