

# 1 **Modelling water quantity and quality for integrated water** 2 **cycle management with the WSIMOD software**

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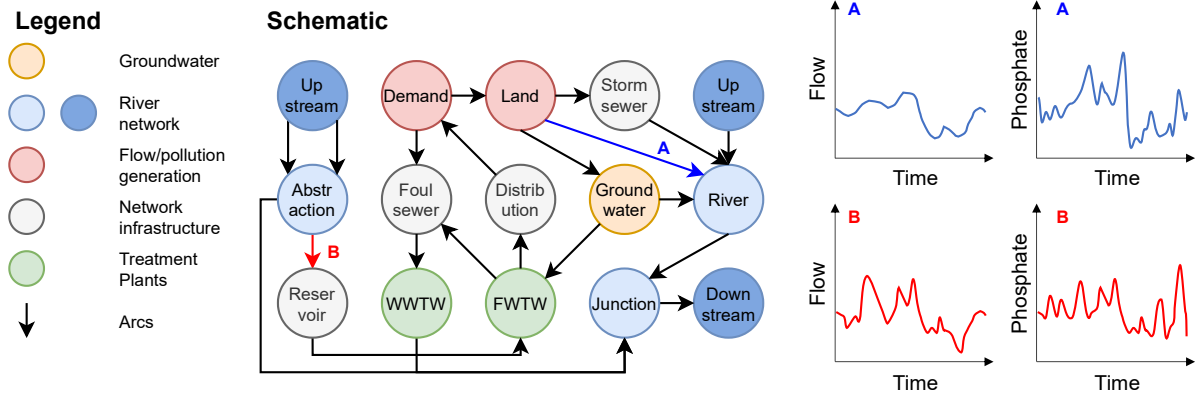
## 7 **Abstract**

8 Problems of water system integration occur when a model's boundaries are too narrow to capture interactions and  
9 feedbacks across the terrestrial water cycle. We propose that integrated water systems models are required to  
10 overcome them, and are necessary to understand emergent system behaviour, to expand model boundaries, to  
11 evaluate interventions, and to ensure simulations reflect stakeholder goals. We present the Water Systems  
12 Integrated Modelling Framework (WSIMOD) software as one such approach and describe its theoretical basis,  
13 covering the node and arc nature of simulations, the integration framework that enables communication between  
14 model elements, and the model orchestration to customise interactions. We highlight data requirements for  
15 creating such a model and the potential for future development and refinement. WSIMOD offers a flexible and  
16 powerful approach to represent water systems, and we hope it will encourage further research and application into  
17 using model integration towards achieving sustainable and resilient water management.

## 18 **Plain Language Summary**

19 Water management is challenging when models do not capture the entire water cycle. We propose that use of  
20 integrated models will facilitate better management and improve understanding. Thus, we introduce a software  
21 tool designed for this task. We discuss its foundation, how it simulates water system components and their  
22 interactions, and its customization. We provide a flexible way to represent water systems, and we hope it will  
23 inspire more research and practical applications for sustainable water management.  
24

25 **Graphical Abstract**



26

27 **1 Introduction**

28 Water fluxes and their pollution concentration are influenced by the interactions and management of all  
 29 components that make up the human-altered terrestrial water cycle, including but not limited to hydrological and  
 30 groundwater catchments, agriculture, upstream rivers, reservoirs, freshwater treatment, distribution networks,  
 31 residential and non-residential water consumers, foul sewers, urban drainage, storm sewers, and wastewater  
 32 treatment works, (and the many physical and operational processes within each component). The importance of  
 33 this interconnectedness is most evident in rivers, which reflect the overall condition of the catchment system since  
 34 they aggregate behaviour over such large areas (Dobson and Mijic 2020; Kirchner 2009). Hydrological  
 35 catchments are rarely dominated by the behaviour of any specific component or any individual stakeholder’s  
 36 decisions. For example, 60% of English hydrological catchments that do not achieve a ‘good’ status in the Water  
 37 Framework Directive (WFD) do so because of multiple different pollution sources, with wastewater infrastructure  
 38 and agriculture being the most prevalent drivers, each affecting 50% of hydrological catchments. (Environment  
 39 Agency 2020b). The implication is that a modelled representation focussing on any individual component is  
 40 unlikely to give accurate estimates of impacts beyond that component (Beven 2007; Blair et al. 2019; Dobson,  
 41 Wagener, and Pianosi 2019; Schmitt and Huber 2006). Furthermore, estimates within a subsystem representation  
 42 may be inaccurate if sensible boundary conditions cannot be defined, something that water managers are highly  
 43 sensitive to (Höllermann and Evers 2017).

44 We briefly pose a variety of questions to further motivate these challenges, some of which are developed  
 45 throughout the paper. Can changes in agriculture fertilising behaviour offset the increased sewage due population  
 46 changes downstream (Liu, Dobson, and Mijic 2022)? Will water efficient appliances concentrate household  
 47 sewage to the extent that river water quality is worsened (Mott MacDonald 2023)? Will wastewater reuse worsen  
 48 low flow conditions in rivers downstream of the wastewater treatment plant where it is implemented (Mott  
 49 MacDonald 2023)? Can upstream water supply river abstractions be strategically reduced on sewer overflow days  
 50 to dilute the spill without compromising supply reliability (Dobson and Mijic 2020)? Will increasing prevalence  
 51 of ‘working from home’ change where sewage is produced to have knock-on impacts on wastewater infrastructure  
 52 and rivers (Dobson et al. 2021)? In a multi-polluter hydrological catchment, can combinations of pollution  
 53 reduction achieve a target water quality at a downstream checkpoint (Liu, Dobson, and Mijic 2023b)? For each

54 of these questions, there will be water systems where these questions are true, and cases where they are not. We  
55 term these problems of water systems integration, and it follows that understanding the terrestrial water cycle  
56 ('water cycle' hereafter) as a whole is needed to address them. Models that take such an approach will better  
57 capture component boundaries and the wider impacts of stakeholder decisions, ultimately enabling more accurate  
58 representations of water quality in rivers, which is essential to effectively manage, for example, water supply  
59 (Mortazavi-Naeini et al. 2019) and biodiversity (Dobson, Barry, et al. 2022).

60 In this paper we introduce the theoretical underpinning behind a novel method for modelling integrated water  
61 systems to address these challenges. Firstly, the need for integrated water cycle simulation models is explained,  
62 including their current coverage. The importance of parsimonious representations within an integrated model is  
63 then discussed, along with the methods used to achieve integration.

64 The environmental modelling research community has responded to problems of water systems integration  
65 primarily through computer simulation models (Bach et al. 2014; Best et al. 2011; Douglas-Mankin, Srinivasan,  
66 and Arnold 2010; Rauch et al. 2017; Tscheikner-Gratl et al. 2019; Whitehead, Wilson, and Butterfield 1998). We  
67 distinguish an integrated water system modelling approach from a system dynamics approach (see Zomorodian  
68 et al., (2018)) by further specifying that component representations must have a physical basis, which is needed  
69 to link observational data to model behaviour and to capture interventions (e.g., new infrastructure or changes to  
70 operations). We define integrated water system models as those which link component representations to capture  
71 and understand the complex interactions and feedbacks that occur between components. We categorise the four  
72 key goals of these models: (1) to understand which fundamental processes drive emergent behaviour at a whole-  
73 water system scale; (2) to avoid simulation inaccuracies caused by narrow boundary conditions; (3) to test  
74 interventions to the physical system or operational behaviour in order to understand their water cycle wide impacts  
75 or interactions and (4) to capture impacts that align more closely with desired water system outcomes, in addition  
76 to performance indicators of individual components. For example, in-river pollutant concentration is a better  
77 indicator of wastewater system performance than the more typically monitored number of sewer spills (Giakoumis  
78 and Voulvoulis 2023).

79 In addressing problems of water systems integration, existing modelling approaches have made significant  
80 progress. Bach et al. (2014) set out a comprehensive typology for integrated urban water systems modelling.  
81 However, among the reviewed models, only CityDrain3 (Burger et al. 2016), WEST (Vanhooren et al. 2003), and  
82 SIMBA (IFAK 2007) can represent receiving water bodies (i.e., rivers), which is where the importance of an  
83 integrated representation is most pronounced. Furthermore, due to the urban focus of these models, the ability to  
84 simulate pollution concentrations in receiving waters impacted by upstream hydrological catchments is highly  
85 limited yet is central to quantifying in-river impacts (Liu et al. 2022). A more recent effort to characterize  
86 integrated water systems modelling places importance on in-river conditions (Tscheikner-Gratl et al. 2019) and  
87 present a comprehensive review of urban and rural water cycles and their impacts on rivers. However, the  
88 reviewed modelling approaches omit some key factors: the importance of water resources infrastructure, which  
89 play a significant role in concentrating pollution during low flows if abstractions take place; the relevance of  
90 groundwater, which provides baseflow to dilute pollution during critical low flow periods; and consideration of  
91 agricultural processes and associated pollution that results from them, which is a critical source of water pollution  
92 worldwide (Mateo-Sagasta et al. 2017; Tang et al. 2021), and the second most common hydrological catchment  
93 pollution source in England (Environment Agency 2020b). Integrated models that capture groundwater and

94 agricultural processes are present in the modelling literature, such as INCA (Whitehead et al. 1998) and HYPE  
95 (Lindström et al. 2010), however, in contrast, these are limited by their ability to capture urban systems. Thus,  
96 while water systems integration is well-served from a rural or urban modelling perspective, we identify that there  
97 is not yet an approach that offers a self-contained representation to capture all key processes required to model in-  
98 river water quality at a whole-water cycle scale.

99 A further critical factor in creating an integrated water systems model is how components are represented. In  
100 general, current approaches have favoured identifying pre-existing detailed component representations which are  
101 then integrated (Schmitt and Huber 2006). For example, DAnCE4Water (Rauch et al. 2017), which is the most  
102 comprehensive application to date, includes high resolution and sophisticated models for a wide range of urban  
103 components. However, as more and more components are captured by integrated modelling, it becomes  
104 increasingly difficult parameterising such detailed models. Simply combining separately calibrated models  
105 provides no guarantee of performance as a whole (Lee 1973). Meanwhile, integrated models typically have many  
106 parameters that may compensate each other, thus making calibration a challenging and risky process (Voinov and  
107 Shugart 2013). An alternative approach is to forego calibration altogether by adopting parsimonious models with  
108 fewer parameters and ideally deriving those parameters from best available data (Dobson et al. 2021). Although  
109 complicated modelling approaches are needed for tasks such as design, these approaches are also more difficult  
110 to apply widely and thus may hinder the goals of integrated water systems modelling. For example, building  
111 scientific understanding requires repeated testing of an approach in various locations, and customising the model  
112 to match local conditions is essential when representing interventions. Therefore, a modelling approach that can  
113 be easily deployed on a wider scale is of significant benefit to problems of water systems integration.

114 To ensure that a range of water system configurations can be accommodated, flexibility or customisability must  
115 be incorporated into the approach for integration. Integration approaches vary broadly between tightly coupled  
116 and loosely coupled. In a tightly coupled approach equations and interactions are pre-defined to create a self-  
117 contained integrated representation, such as with JULES (Best et al. 2011) or INCA (Whitehead et al. 1998).  
118 While, in a loosely coupled approach, component representations are self-contained, and the integration occurs  
119 by facilitating their interactions, filling the role as a message passing interface, such as with OpenMI (Harpham,  
120 Hughes, and Moore 2019) and DAnCE4Water (Rauch et al. 2017). Belete et al. (2017) describe the arrangement  
121 of components and their interactions as integrated model orchestration, highlighting that different orchestrations  
122 are suitable for different applications. While looser coupling provides greater control over orchestration, and thus  
123 greater ability to customise and capture a wide variety of systems, it also creates a higher user burden to set up  
124 and understand many subsystems, considered to be a key barrier to the uptake of such approaches (Zomorodian  
125 et al. 2018). Conversely, a tightly coupled model that represents the same components as a loosely coupled one  
126 may be easier to set up but typically offers less control over orchestration. In the middle ground is an integrated  
127 representation that gives flexibility around orchestration but comes with self-contained components that do not  
128 need to be onerously setup by a user, such as the CityDrain3 software for modelling urban drainage systems  
129 (Burger et al. 2016). We propose that this middle ground is the most beneficial for a modeller and believe that  
130 such an approach to integration is the most productive avenue towards creating highly flexible, user-friendly  
131 models of the integrated water cycle.

132 The concepts introduced above suggest that, for many problems of water systems integration, capturing a broad  
133 representation of the water cycle and interactions between its components is equally important as detailed

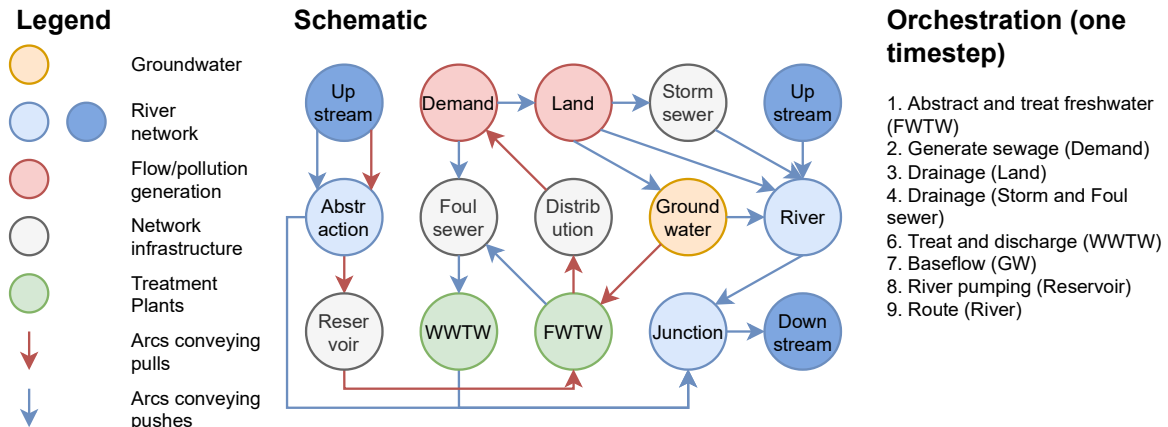
134 component representations. We have created a tool to implement this modelling philosophy, the Water Systems  
135 Integrated Modelling framework (WSIMOD), which is an open-source Python package for flexible and  
136 customisable simulations of the water cycle that treats the physical components of the water cycle as nodes  
137 connected by arcs that convey water and pollutant flux between them. The software source code and online  
138 tutorials are published by Dobson, Liu, and Mijic (2023a), in contrast, this paper presents WSIMOD's theoretical  
139 underpinning with a discussion on model setup and of integrated water system modelling in general. To address  
140 the difficulties in application associated with integrated modelling mentioned above, WSIMOD contains a library  
141 of built-in component representations covering a more complete water cycle coverage than any identified  
142 integrated models, and a default but customisable orchestration deemed to be suitable for many catchments and  
143 regional water systems coordination. Where possible these representations are based on parsimonious and peer-  
144 reviewed models. Extensive model documentation with worked examples is provided online (Dobson, Liu, and  
145 Mijic 2023b), enabling users to gain confidence and become familiar with using WSIMOD.

## 146 2 WSIMOD

147 WSIMOD is an integrated modelling framework that provides ready-to-use objects (nodes, arcs, water stores, and  
148 model orchestration) that are suitable for a wide range of water systems and described in greater detail in the  
149 following sections. However, WSIMOD is not intended to be a one-size-fits-all solution, indeed, the ubiquity of  
150 non-textbook water systems led us to create a more customisable modelling approach in the first place. This paper  
151 describes the theory behind WSIMOD in general and user-friendly terms, avoiding the use of equations and  
152 technical details, while further documentation can be found online (Dobson et al. 2023b). The WSIMOD  
153 framework is implemented in Python 3, which is widely practiced in the environmental modelling community and  
154 facilitates quick setup and easy customisation. WSIMOD is the combined effort of many studies conducted as  
155 part of the CAMELLIA (Community Management for a Liveable London) project  
156 (<https://www.camelliawater.org/>), which are linked to relevant sections of the model description to highlight the  
157 range of possible applications.

158 An example WSIMOD model schematic is shown in Figure 1. The arrangement is based on the city of Oxford,  
159 UK, and is used in a demonstration in Section 3. Oxford was selected because its water cycle is highly integrated  
160 but also self-contained. We first use this schematic to explain the generic water cycle implemented in WSIMOD,  
161 however, we note that this is customisable, see Sections 2.2.3 and 2.3. A given timestep, see 'Orchestration' in  
162 Figure 1, begins at freshwater treatment works (FWTW), which pump water from various sources (groundwater  
163 and a reservoir in this example), treat that water to fill their service reservoirs, and send sludge to the Foul sewer.  
164 Demand nodes (i.e., population and other water users) then generate water demand, retrieve water from the  
165 FWTW service reservoirs via a Distribution node, satisfy any garden irrigation demands (the link between  
166 Demand and Land) and discharge wastewater to foul sewers. The hydrological and agricultural processes are then  
167 run within the Land node, these are substantial and have a tutorial in the online documentation  
168 ([https://imperialcollegelondon.github.io/wsi/demo/scripts/land\\_demo/](https://imperialcollegelondon.github.io/wsi/demo/scripts/land_demo/), accessed 2024-04-08), in short, they track  
169 crop growth and calendars, apply fertilisers, nutrient cycling and erosion, generate impervious runoff (to Storm  
170 sewers), percolation (to Groundwater) and surface/sub-surface flows to the River. Foul and Storm sewer node  
171 processes then run, generating flows to the Wastewater Treatment Works (WWTW) and into the River

172 respectively. Baseflows are then calculated based on conditions in the Groundwater node, and reservoir  
 173 abstractions are made. The timestep completes by routing upstream to downstream along the river network.



174  
 175 Figure 1: An example WSIMOD model for Oxford, UK. Orchestration is shown demonstrating the high-level  
 176 functions called for each timestep. Nodes are shown as circles and arcs as arrows. WWTW stands for Wastewater  
 177 Treatment Works and FWTW stands for Freshwater Treatment Works.

178 The aggregated nature of the components depicted, for example, an individual node represents the entire storm  
 179 sewer network, is a common feature of many WSIMOD models (but not always, see Dobson et al., (2022)). A  
 180 general philosophy that drives WSIMOD, as anticipated in the introduction, is that it is easier to introduce  
 181 complexity into an already integrated model rather than to integrate separate complex models. To achieve such  
 182 customisation and introduction of new behaviours, WSIMOD uses object-oriented programming (OOP), which  
 183 classifies components by common attributes and behaviours (classes). All objects in WSIMOD are a subclass of  
 184 WSIObj, which predefines efficient arithmetic operations for water quality and volume, however users will  
 185 typically instead interact with the subclasses described in the following sections. Additionally, users may  
 186 customise a model's high-level control over how interactions take place within a timestep, or the model's  
 187 orchestration (Belete et al. 2017), which is a unique feature of WSIMOD and further developed in Section 2.3.  
 188 Thus, while Figure 1 depicts one possible arrangement and selection of nodes and arcs, a wide variety of water  
 189 systems can be represented. Section 3 presents this Oxford WSIMOD model in its simplest integrated form and  
 190 demonstrates how to introduce complexity into such a model (specifically in this example, to introduce more  
 191 complex water resources behaviour). If readers wish to be involved with the development of WSIMOD, identify  
 192 bugs, or require further clarification in terms of documentation, we recommend the viewing Contributing section  
 193 of the main repository page (<https://github.com/ImperialCollegeLondon/wsi>, accessed 2024-04-08).

194 **2.1 Nodes represent water cycle components**

195 Physical representations of the different components in the water cycle are typically implemented as WSIMOD  
 196 nodes, Figure 1. In the software implementation, all nodes are instances of the Node class or its subclasses.  
 197 Formulation of components as nodes using OOP draws heavily on the CityDrain3 software (Burger et al. 2016).  
 198 Our generic definition allows nodes to represent diverse entities, for example, a collection of manholes  
 199 representing a region of sewer network or individual manholes that can be connected to represent a sewer network;

200 as demonstrated in (Dobson, Watson-Hill, et al. 2022). In this section we describe the Node class, summarise the  
201 existing node subclasses currently implemented in WSIMOD, and describe how to customise them.

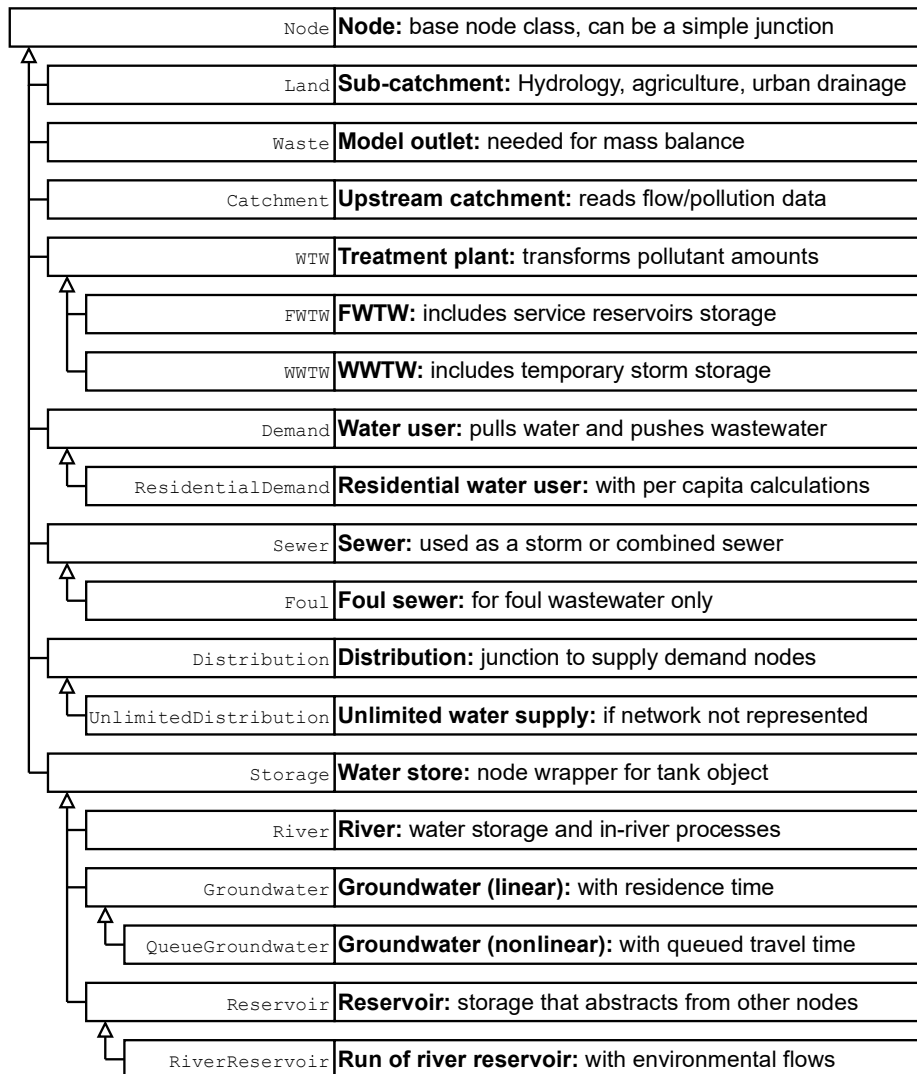
### 202 **2.1.1 The Node class**

203 The Node class in WSIMOD is a generic class that is expected to be the parent of any component represented.  
204 The class captures three key behaviours. Firstly, it predefines defaults for much of the interaction functionality  
205 later described in Section 2.2. Secondly, it contains a variety of useful functions to interact with other nodes via  
206 arcs (see Section 2.2.1). Thirdly, it enables mass balance checking (see Section 2.3.2). For these reasons, physical  
207 components should generally be implemented in WSIMOD as a subclass or variation of a Node, even if the  
208 computational implementation is simply a wrapper for another, pre-existing, model. Although the base Node class  
209 does not implement any physical processes, it may serve as a junction for branches or convergences in the water  
210 system, such as river bifurcations and confluences.

### 211 **2.1.2 Node subclasses**

212 To ensure WSIMOD is as easy to be implemented as possible, a variety of water cycle components have been  
213 developed in Python as Node subclasses. We summarise these components in Figure 2 and the Component Library  
214 section in the online documentation, for full details we recommend viewing the API reference and ‘Key  
215 assumptions’ section of Node subclasses in the online documentation, as well as various tutorials on their use  
216 (Dobson et al. 2023b). As anticipated in the introduction, these components are designed for parsimony and can  
217 be instantiated with as few parameters as possible, thus minimising data requirements and maximising utility. We  
218 note that this summary is up to date as of time of writing, however the Imperial College London Water Systems  
219 Integration research group will be continually upgrading and adding new functionality to WSIMOD.

220



221  
 222 **Figure 2: An inheritance diagram of components implemented as nodes in WSIMOD, arrows indicate that a node is a**  
 223 **subclass of another node. The courier text is the name of the node in WSIMOD.**

224 Many node types include water stores, which are sufficiently prevalent in water systems to warrant their own class  
 225 in WSIMOD, referred to as a Tank, with further details provided in A1.1 Tank object to generalise water stores.  
 226 It's important to note that a tank is not a node, but a node may have a tank. The number of tanks a node subclass  
 227 may have varies depending on what it represents, with some having none (e.g., Demand), one (e.g., Reservoir),  
 228 or multiple tanks (e.g., Land). If a user requires a tank that can act as a node, they can utilize the Storage class to  
 229 achieve this functionality.

### 230 2.1.3 Customising nodes

231 We do not expect that the provided WSIMOD components will be sufficient to cover every water system or be  
 232 suitable in cases where detailed representations are necessary. Where possible, users should consider customising  
 233 model orchestration (Section 2.3) or component interactions (Section 2.2.3) to represent these cases. However,  
 234 new physical subsystems may also be represented based on those defined in Figure 2. The variety of techniques  
 235 to customise subclass behaviour in OOP are outside the scope of this paper but have been discussed extensively  
 236 elsewhere (Gamma et al. 1995). In general, these techniques aim to avoid duplication of effort, because most

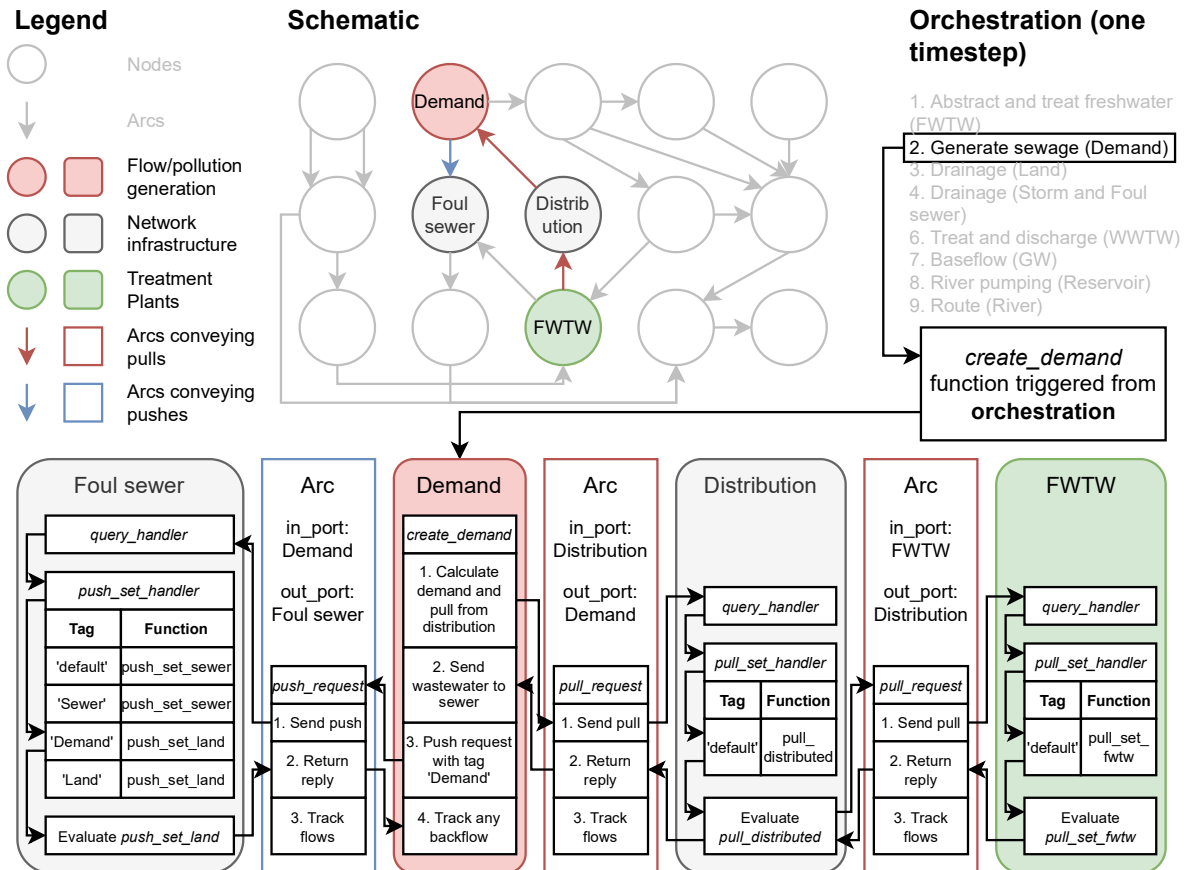


237 functionality will be predefined in an existing class that can maintain specific interactions between existing  
238 subclasses. We enthusiastically encourage users to contribute to the open-source package to create a better  
239 software for the wider environmental modelling community, though request that contributors read our online  
240 guidelines before doing so to avoid highly duplicative and bloated components that are likely to confuse others.  
241 If a pre-existing model aims to be integrated and it would be too significant a programming effort to reimplement  
242 under this philosophy, then we would request that wrappers are used to treat the model as a node and interface  
243 with the existing software.

244 To highlight the flexibility offered by WSIMOD, we briefly discuss some examples of complex behaviour being  
245 captured through node customisation. In Dobson et al., (2021), demand nodes were assigned time varying  
246 population, water use behaviour, and pollutant generation to capture changing commuter patterns in London that  
247 resulted from the COVID-19 pandemic. In Liu et al., (2023a), the hydrological processes in land nodes were  
248 customised in a variety of ways to represent nature-based solutions. We note that these examples are provided  
249 with open-source code but are not included in the WSIMOD repository because they have significant data  
250 requirements to set up that are context specific, and thus are unlikely to be generalisable to a wide range of cases.  
251 In the software documentation we also provide a how-to guide for node customisation (Dobson et al. 2023b).

## 252 **2.2 Integration framework**

253 The WSIMOD integration framework facilitates interactions between nodes by serving as a message passing  
254 interface that transfers information relating to water quality and quantity. It was developed in an application to  
255 London's water cycle at a wastewater catchment scale (Dobson et al. 2021). It draws significantly on the OpenMI  
256 (Harpham et al. 2019) and Open MPI (Graham et al. 2006) interfaces but has been tailored to water systems by  
257 providing a variety of built-in behaviours. The integration framework consists of three key concepts: arcs, pushes  
258 and pulls, and requests and checks (demonstrated in Figure 3 and described in the following sections). Arcs are a  
259 class that facilitate interactions between nodes but can also represent physical entities (e.g., pipes). Arcs convey  
260 both water quality and quantity fluxes, which are discretised and packaged together, based on concepts from  
261 CityDrain3 (Burger et al. 2016). Pushes and pulls differentiate between the directionality of an interaction.  
262 Requests and checks differentiate between information passing that simulates the movement of water (requests)  
263 and that which does not (checks). While we do not recommend changing the integration framework itself, we  
264 provide a generic method to accommodate a wider variety of interactions in Section 2.2.3.



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**Figure 3: An example of the WSIMOD integration framework, illustrated through the automatic behaviour triggered when the `create_demand` function is called by a Demand node during orchestration. The node pulls water via an arc from a distribution network and pushes foul wastewater via an arc to a sewer. The figure further illustrates the use of handlers and tags to customise interactions between nodes. Italicized text indicates that it is a function. Nodes are shown as circles or rounded squares, while arcs are shown as coloured arrows or sharp coloured squares.**

### 271 2.2.1 The Arc class and fluxes

272 Arcs are a class to establish connections between nodes. They transmit all message passing and track fluxes when  
 273 requests are made. Arcs can have a capacity property that limits the flow in a given timestep, which can also be  
 274 customised to be dynamically calculated, for example, to implement Manning’s equation along pipes (Dobson,  
 275 Watson-Hill, et al. 2022). Additionally, if multiple arcs are linked to a single node, they can be assigned a  
 276 preference attribute. This enables the node to prioritize certain arcs over others. For example, a sewer node may  
 277 connect to a wastewater treatment plant (WWTW) and to a river via a sewer spill. In this case, the spill arc could  
 278 be assigned a low preference to ensure that it is only utilized if the WWTW cannot accept any water.  
 279 Fluxes in WSIMOD are described in discrete packages called Volume-Quality Information Packages (VQIP). A  
 280 VQIP is a dictionary that contains entries for volume and all simulated pollutants. Calculations in WSIMOD are  
 281 typically performed on VQIPs rather than simply flows, thus ensuring simulation of both water quantity and  
 282 quality. The core parent class (WSIObj) of all WSIMOD classes provides functions to perform basic operations  
 283 with VQIPs in place of the normal arithmetical operations that would typically be only performed on flows or  
 284 volumes. VQIPs track water quality as mass rather than as concentration to accommodate cases where pollutants  
 285 are being moved with no associated water quantity, except for non-additive variables such as temperature or pH.

286 WSIMOD refers to anything tracked in a VQIP that is not water volume as a pollutant, however this should be  
287 interpreted as a water quality constituent, and does not imply that everything simulated (e.g., temperature or  
288 dissolved oxygen) is a pollutant from an environmental/management perspective.

289 WSIMOD can simulate any number of pollutants in a mass balance approach, also referred to as conservatively,  
290 provided their sources into the water cycle can be identified and quantified. However, bio-chemical changes for  
291 pollutant decay can also be represented, see Appendix 1, A1.2 Non-flux pollutant changes. Other more  
292 complicated pollutant transformation can be captured on a case-by-case basis, for example to capture nutrient  
293 cycling in the soil pool, using on equations from Liu et al., (2022).

294 A further consideration commonly required in water systems is that of travel time of water, which requires its own  
295 specific implementation due to the discrete nature of flux and VQIPs in WSIMOD, further details are provided in  
296 A1.3 Travel time of water.

## 297 **2.2.2 Types of interactions: pushes/pulls and requests/checks**

298 In order for a simulation to occur in a non-tightly coupled integrated representation, something must trigger the  
299 interactions that are conveyed via arcs. High-level controls that govern these behaviours are described as model  
300 orchestration (see Section 2.3), however a key benefit to using this integration framework is that the user does not  
301 have to predefine all possible interactions in advance. Because information transmitted by arcs automatically  
302 triggers further information transmission in connected nodes, a user may customise their nodes, arcs, and  
303 orchestration, without onerously updating every possible interaction that may take place in the model, as  
304 visualised in Figure 3.

305 To represent a wide variety of behaviours, WSIMOD categorises interactions based on directionality of intent and  
306 whether they represent flux or not. Directionality of intent refers to either cases when a node has water that must  
307 be sent somewhere, called a push, or when a node needs water from somewhere, called a pull. Interactions between  
308 nodes, whether pushes or pulls, may convey flux of water and pollutants, and simulate the movement of water, or  
309 they may convey non-flux information necessary for achieving realistic simulations. An interaction conveying  
310 flux is referred to as a request, while a non-flux interaction is a check.

311 Pushes occur when a node needs to discharge water. For example, when a wastewater treatment works (WWTW)  
312 must discharge effluent to a river, or a hydrological catchment must discharge runoff downstream. In general,  
313 push scenarios are more common in water systems because water travels from upstream to downstream.  
314 Meanwhile, pulls occur when a node requires water. For example, when a farmer pumps water from a borehole  
315 to fill their irrigation reservoir. Pulls typically represent human-related effort to move water in a non-natural way.  
316 A request occurs when a node intends to push or pull a certain amount of water to or from another node, regardless  
317 of the connected node's current state. For example, in the pull case, a demand node will intend to satisfy its entire  
318 water needs by pulling water from the distribution network, and this intention is independent of the availability of  
319 water in the distribution network. For example, in the push case, a demand node will always intend to send the  
320 entire volume of its foul water to a sewer system, even if the sewer cannot accommodate the full amount. All  
321 water flux in WSIMOD is ultimately simulated by requests, however most cases are not as straightforward as the  
322 above examples, and nodes often require additional information about the state of giving or receiving nodes to  
323 calculate their requests. Interactions passing such non-flux information are referred to as checks.

324 Generically a check is any kind of non-flux information passing between nodes, it enables a node to use the state  
 325 of the nodes that it interacts with to calculate its requests. For example, when a freshwater treatment works  
 326 (FWTW) can draw water from multiple viable reservoirs collectively containing more water than needs to be  
 327 treated, a calculation is required before requesting water. The FWTW will send checks to the connected reservoirs  
 328 to determine their available water capacity and calculate the appropriate ratio to satisfy its treatment demand. Only  
 329 then will the FWTW send requests for the required water.

### 330 2.2.3 Default and customised interactions

331 During simulation, a node needs to make responses when it receives a push/pull request/check, which we term as  
 332 a reply. We formulate four types of predefined replies that are widely observed in node interactions in the water  
 333 systems. These default replies allow nodes to interpret the responses of requests/checks sent to the interacting  
 334 nodes without the need to understand their detailed behaviour. These defaults are set out in Table 1 with examples.

335 **Table 1: The default reply from each kind of component interaction with an example.**

	Push		Pull	
	Reply	Example	Reply	Example
Request	Amount not received	A sewer sends a push request to a WWTW. The WWTW calculates the available capacity, updates its state variables to represent the increased throughput, and replies with how much water from the request that could not be received.	Amount sent	A FWTW sends a pull request to a reservoir. The reservoir calculates how much of the request can be met, updates its state variables to decrease the current volume, and replies with the amount of water abstracted.
Check	Maximum volume available to push	A sewer sends two push checks to two downstream sewers (required to calculate what proportion to discharge water to them). The two sewers reply with total amount of water that they can each receive.	Maximum volume available to pull	A FWTW sends two pull checks to two reservoirs (required to determine what proportion to pull water from them). The reservoirs reply with each of their current abstractable volumes.

336 While these default interactions are likely to accommodate much of the information passing required to simulate  
 337 an integrated water cycle, further customisation may be necessary to allow specific nodes to respond differently  
 338 to others. For example, a sewer node might respond differently to a push request from another sewer than from a  
 339 land node because sewer to sewer travel time may be calculated differently than land runoff to sewer travel time.  
 340 Furthermore, the default check behaviour which transmits capacity information may not be sufficient for a  
 341 component to calculate where to push/pull water. For example, due to the importance of head in determining flow,  
 342 the amount of water that a floodplain can discharge would not be based on the receiving river's capacity but  
 343 instead its head (Liu et al. 2023a).

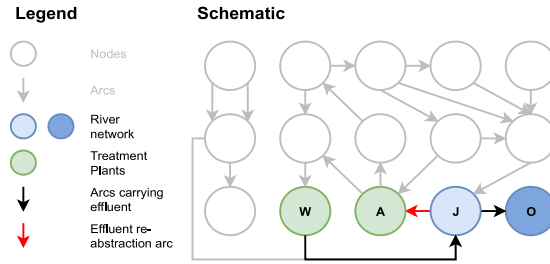
344 To define what a node does in reply to an interaction, all WSIMOD interactions pass through ‘handlers’ that are  
345 associated with a ‘tag’. A handler is a Python dictionary belonging to a node, and the tag is a key to the handler  
346 that determines which function is called during a given interaction (see Figure 3). All nodes must have handlers  
347 which contain a ‘default’ tag, thus enabling any node to interact with any other node. However, additional tags  
348 and replies can be added to enable different behaviours based on the type of node that is interacting with it. An  
349 example interaction for a Demand node draining to a foul Sewer node is given in Figure 3. The generation of  
350 wastewater in the Demand node is triggered by orchestration and sent as a push request, via an arc, to a sewer  
351 node, with a tag ‘Demand’ to indicate how the sewer node should reply. The sewer node has predefined behaviour  
352 for tags ‘default’, ‘Sewer’, ‘Demand’, and ‘Land’, with different functions associated with these tags. The sewer  
353 node uses the handler to identify that it should evaluate the ‘push\_set\_land’ function (which calculates the travel  
354 time through the sewer using the same equation as is used for calculating travel time of runoff arriving from a  
355 land node) and returns the reply via the arc. The base Node predefines simple default handlers that enable it to  
356 convey interactions to connected nodes, for example, a push request to a Node will trigger push requests to  
357 connected nodes, in this way the Node can behave as a simple junction. A how-to guide to explain this behaviour  
358 in greater detail, with examples, is available in the documentation (Dobson et al. 2023b).

### 359 **2.3 Orchestration to manage and enable simulation**

360 As defined by Belete et al. (2017), orchestration is how nodes and their interactions are managed and enabled.  
361 The default steps that we consider to be important to include in orchestration for many water systems are shown  
362 in Figure 1. In integration frameworks such as OpenMI, all interactions in a simulation occur by nodes triggering  
363 pulls to other nodes, orchestrated by a single external pull each timestep (Harpham et al. 2019). Meanwhile, in  
364 WSIMOD, a finer level of control of orchestration is given to a user. We argue that the WSIMOD approach is  
365 better at capturing complex within-timestep behaviours.

366 In Figure 4, building on the system shown in Figure 1, we provide a motivating example of how it is more efficient  
367 to customise the behaviour of water systems by customising orchestration, rather than customising/adding nodes  
368 or arcs, as would be required in other integration frameworks. We consider a common case for an integrated water  
369 system of requiring the simulation to perform downstream re-abstraction of wastewater effluent, visualised by the  
370 red arc in Figure 4. Wastewater re-abstraction (water from W reaching A) requires available wastewater at W to  
371 be matched against demand for re-abstracted wastewater at A, both calculations involving the interconnected  
372 node, J.

(A)



(B) Pull driven system

Orchestration	Node	Automatic trigger	Component customisation
External pull	-	Pull from A and O	Trigger A before O
↳ A	J	Pull from J	
↳ J	W	Pull from W	Create custom exchange item conveying only the requested abstraction from A
↳ W		(Pull) Simulate flow to J	
↳ J		(Pull) Simulate flow to A	
↳ O	J	Pull from J	
↳ J	W	Pull from W	
↳ W		(Pull) Simulate flow to J	
↳ J		(Pull) Simulate flow to O	

(C) Push-pull custom orchestration system

Orchestration	Node	Automatic trigger	Component customisation
Calculate available wastewater	W		
Satisfy abstractions	A	Pull from J	
↳ J	W	Pull from W	
↳ W		(Pull) Simulate flow to J	
↳ J		(Pull) Simulate flow to A	
Discharge remaining water	W	(Push) Simulate flow to J	
↳ J		(Push) Simulate flow to O	

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**Figure 4: (A) an example water system based on the Oxford example from Figure 1 that includes re-abstracted of treated effluent (red arc). (B) the model steps to achieve re-abstracted of effluent when using a pull-driven integration framework, of the kind used in OpenMI, without customisable orchestration, (C) and when using a push-driven integration framework, of the kind used in WSIMOD, with customisable orchestration.**

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To capture such an interaction without customising orchestration and using only pulls, as would be required with an OpenMI approach (Figure 4 (B)), customisation of J would need to specify whether the pull is originating from A or O, and this information would need to be conveyed to W. This is because, if the pull is intended to route the system for that timestep (i.e., it is originating from O via J), then W should release all its effluent, while if it is intending to satisfy the demand (i.e., it is originating from A via J), it should only release enough to meet A's requirements. Additionally, the external pull that initially triggers all interactions must be customised to pull from A before O, so that W does not fully route before abstractions can be made.

Instead, a user with a broad overview of the water cycle may more easily accommodate such behaviour if given high-level control over the model orchestration (Figure 4 (C)). During a timestep the orchestration can specify that W first calculates its treated effluent without discharging it into a receiving river, abstractions are triggered at A, which draws water from W via J, and then W discharges any remaining effluent into its receiving water. This flexibility in orchestration enables representations of a wide variety of water systems while minimizing changes to the behaviour of underlying components.

### 391 **2.3.1 The Model class**

392 A Model class contains all nodes, arcs and forcing data, and provides a default orchestration that we expect can  
393 represent a wide variety of water systems. Whether the WSIMOD default orchestration is used or whether an  
394 entirely new orchestration is defined, we recommend the use of a Model class for a variety of reasons. Firstly,  
395 built-in load and save functionality enables easy sharing and editing of a specific model. Secondly, it contains a  
396 run function that can carry out orchestration, perform mass balance checking (see following section), store  
397 simulation results, and trigger end of timestep functions. Thirdly, it enables easy collection and grouping of nodes  
398 to facilitate orchestration, for example, all WWTWs are referenced by dictionary belonging to the Model class so  
399 that they can easily be triggered during a timestep. A tutorial in use of the Model class is provided in the online  
400 documentation (Dobson et al. 2023b).

### 401 **2.3.2 Software quality control**

402 Due to the integrated nature of the water systems that WSIMOD simulates, it can be easy to introduce errors that  
403 are difficult to spot. We provide extensive unit testing in line with best software development practice, which  
404 enables ensuring that changes to code do not introduce unintended behaviour changes. However, a further  
405 safeguard against this is a unified method for mass balance error checking. Both nodes and arcs have predefined  
406 lists of functions to calculate the total inflows, total outflows and change in any storage. Any newly defined  
407 behaviour for nodes and arcs must then also consider how mass balance checks will be impacted and thus update  
408 the lists of functions associated with inflows/outflows/change in storage.

409 Because the core functionality of WSIMOD that performs mass balance checking is indifferent to units (see  
410 Section 2.3.3), and because some pollutants exist in far smaller quantities than others, it is possible that both  
411 incredibly small and large numbers may be present in VQIPs in the model. Therefore, mass balance checking  
412 compares at the magnitude of the largest value of inflows/outflows/storage for a given pollutant or water volume  
413 for a given model element in each timestep. Any discrepancy that is larger than a user specified value is reported,  
414 although because of floating point accuracy some discrepancies will be unavoidable. We encourage users to  
415 exercise common sense to not chase down incredibly small discrepancies while seeking to understand larger  
416 discrepancies, which are usually indicative of some implementation error. The user control over orchestration  
417 typically makes debugging WSIMOD models easier than most integrated models, as a user may step through each  
418 set of triggers in the orchestration and recheck mass balances until the error occurs.

### 419 **2.3.3 Units and timesteps**

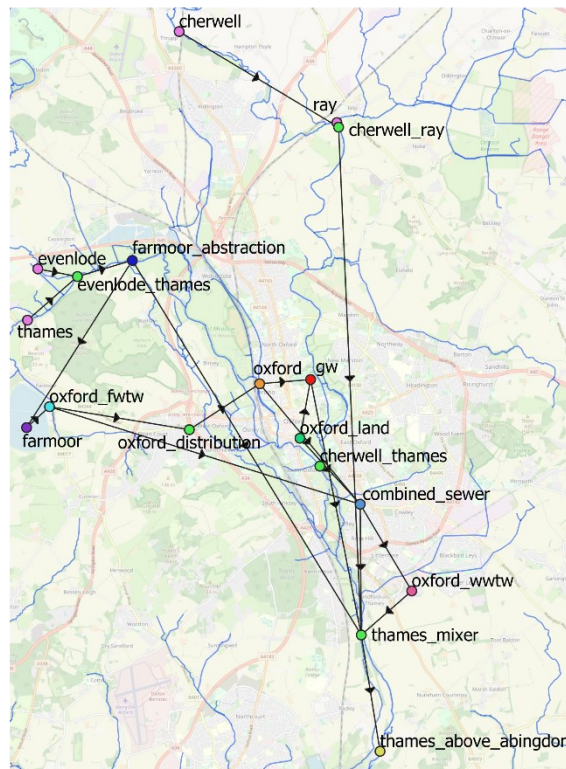
420 While the core of WSIMOD is agnostic of units, many nodes are parameterised and assume input data is in SI  
421 units, thus we recommend use of SI units throughout. If extensive work would be required to re-implement  
422 equations in new units, we recommend converting before and after the calculations from/to SI units.

423 Because the timestep size will vary depending on the application of WSIMOD, nodes do not necessarily make  
424 assumptions about timestep, instead requiring the user to define the timestep that is consistent with their  
425 parameters and input data. This enables significant flexibility in representations that can enable studies  
426 investigating the impact of timestep size on simulations (Dobson, Watson-Hill, et al. 2022). We note that the

427 detailed bio-chemical processes used in agricultural surfaces and rivers assume a daily timestep. In addition, these  
428 processes are developed to focus on pollutants associated with crop nutrient cycles.

### 429 3 Demonstration

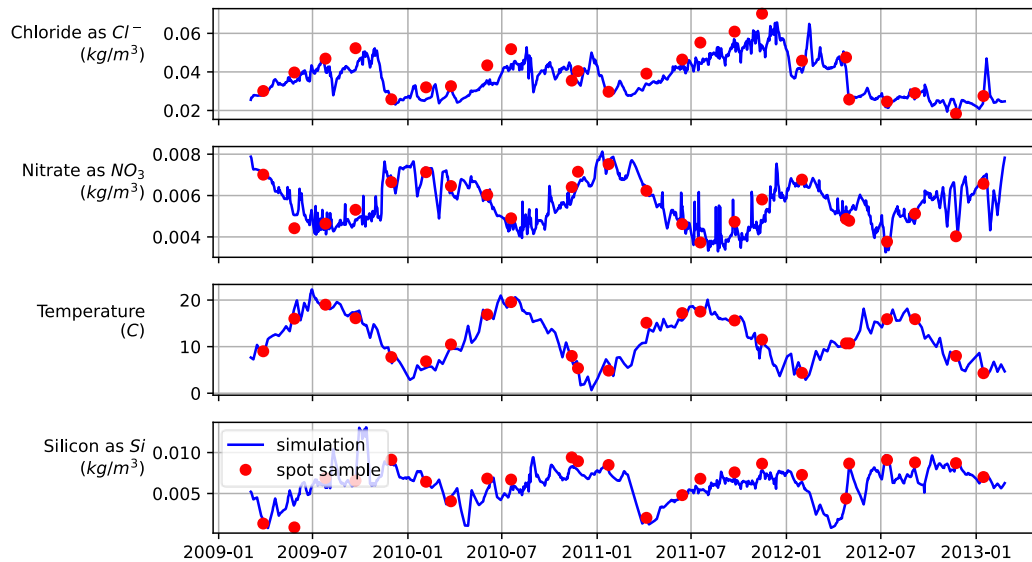
430 Throughout this paper, numerous WSIMOD case studies are referenced. However, to illustrate how WSIMOD  
431 works, we present a demonstration case study. We have chosen a model featured as a tutorial in the online  
432 documentation so that readers will be able to easily reproduce, run and edit it  
433 ([https://imperialcollegelondon.github.io/wsi/demo/scripts/oxford\\_demo/](https://imperialcollegelondon.github.io/wsi/demo/scripts/oxford_demo/), accessed 2024-04-08). The  
434 demonstration covers the area of Oxford, UK, depicted in Figure 5 both as a map and schematic. We highlight  
435 that this simple demonstration and online tutorial is included because of its usefulness in explaining the underlying  
436 functionality and flexibility of WSIMOD, rather than for motivating the complexity and importance of integrated  
437 modelling in general.



438  
439 **Figure 5: Map of the model nodes and arcs over the city of Oxford, map base layer © OpenStreetMap contributors,**  
440 **licensed under the Open Database License (ODbL), rivers data from Open Rivers © 2023 Ordnance Survey Limited.**

441 The model includes four upstream hydrological catchments, a reservoir-based water supply system, and a  
442 combined sewage system, all of which ultimately combine and drain into the River Thames. Because the model  
443 is primarily demonstrative, the parameters given in the model are estimated based on local knowledge. However,  
444 high-resolution (weekly sampling) water quality data are available in the area (Bowes et al. 2018) thus enabling  
445 accurate boundary conditions at upstream hydrological catchments and water quality representations in the model.  
446 In Figure 6 we plot simulations against the weekly sampling, water quality indicators to plot were selected based  
447 on those that had data for all locations for the simulation duration.

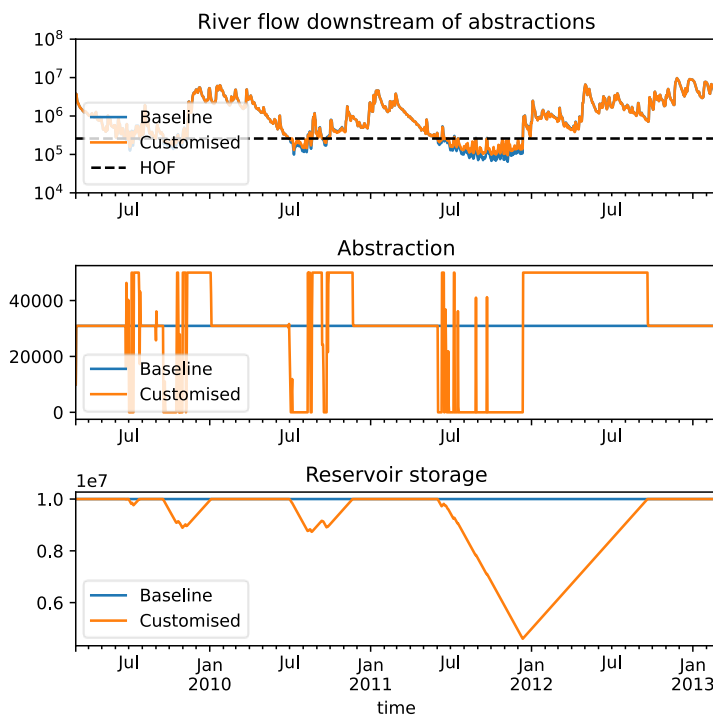




448

449 **Figure 6: Demonstration of simulated levels of chemical water quality indicators against spot samples.**

450 In addition to WSIMOD's capabilities as a general-purpose simulator of hydrological and water quality, it is also  
 451 a valuable tool for management and interventions. The software's design prioritizes customization, making it easy  
 452 to incorporate specific operational preferences which are not typical to capture due to the limited modelling scope  
 453 of most water simulators. For example, in the online customisation guides  
 454 (<https://imperialcollegelondon.github.io/wsi/how-to/>, accessed 2024-04-08), we show how the behaviour of nodes  
 455 and arcs can be altered to accommodate changing abstraction licencing and environmental flow requirements for  
 456 the Oxford case study, with results shown in Figure 7. Liu et al., (2023b), demonstrate how a user can implement  
 457 highly sophisticated water quality management strategies in WSIMOD based on pollution load allocation.



458

459 **Figure 7: Example of implementing a 'hands-off flow' (HOF), which abstractions cannot draw water below, for the**  
 460 **reservoir abstraction node in the Oxford case study model. The 'customised' simulation are with HOF implemented,**  
 461 **and 'baseline' simulation without.**

## 462 **4 Discussion**

### 463 **4.1 Why are integrated water systems models necessary?**

464 We defined four key goals for integrated models of water systems. In this section we will discuss how WSIMOD  
465 helps to meet these goals.

#### 466 **4.1.1 Integration to understand emergent behaviour in water systems**

467 Emergent behaviours in human and environmental systems arise from the interactions between multiple  
468 components and are difficult to predict and understand without addressing the complexity that occurs from this  
469 interconnectedness (Liu et al. 2007). The human-altered water cycle covers a diverse set of components and so  
470 any attempt to understand the fundamental (physical and operational) drivers behind emergent behaviours must  
471 acknowledge the interactions and feedbacks between humans and hydrological processes (Wada et al. 2017).  
472 WSIMOD represents many components within water cycle needed to capture these interactions to reveal and  
473 quantify fundamental processes, for example, that in-river water quality during wetter periods is driven by  
474 agricultural processes, and during drier periods by urban processes (Liu et al. 2022). Without integrated modelling,  
475 we risk oversimplifying the system and omitting feedbacks that could have significant implications for water  
476 management decisions (Dobson and Mijic 2020).

477 Understanding and reducing uncertainty behind behaviour in hydrological systems requires intercomparison of  
478 hydrological process representations and thus a flexible modelling framework (Knoben et al. 2019). We suggest  
479 the same is true for the wider water cycle, and so anticipate that the flexible approach to integrated modelling in  
480 WSIMOD is well suited for these purposes, for example, by comparing different assumptions around water  
481 consuming behaviour (Dobson et al. 2021). Furthermore, customisability enables accommodating unconventional  
482 water systems that may stress the assumptions of underlying process representations. This approach can help  
483 identify any weaknesses or gaps in the model and refine our understanding of the both the process in question and  
484 the behaviour of the wider system. For example, the poor simulations of hydraulic structures identified in Dobson  
485 et al. (2022) are a result of reduced accuracy in capturing water head, in turn caused by the discrete modelling  
486 scheme inherent to a non-tightly coupled integrated framework. Identifying this weakness has implications beyond  
487 simulating hydraulic structures and provides guidance for future work.

488 Finally, research that contributes to knowledge must be reproducible, and if it is not then it can hardly be  
489 considered science (Hutton et al. 2016). To ensure that WSIMOD applications are reproducible the software is  
490 open source with a permissive licence, and it is provided with significant documentation and worked examples to  
491 ensure that it is used as intended. This documentation transparently lists the assumptions made for each model  
492 component, both in the source code of that component and in a self-contained library page (Dobson et al. 2023b).  
493 Furthermore, the ability to save the Model class in a self-contained and human readable file enables publishing  
494 WSIMOD applications in an easy to reproduce way.

#### 495 **4.1.2 Integration to expand model boundaries**

496 The WSIMOD modelling framework offers a versatile approach to representing various processes that are  
497 commonly treated as boundary conditions in water cycle models. One example is the agricultural-hydrological

498 Land node, which utilizes the CatchWat model to estimate the pollution and flow in rivers upstream of the model  
499 region (Liu et al. 2022, 2023a). Accounting for upstream pollution and dilution is crucial in accurately assessing  
500 the impact of urban processes on in-river conditions. By incorporating the Land node, WSIMOD provides a means  
501 to capture the typically neglected upstream agricultural system in urban wastewater studies, which is often viewed  
502 as a boundary. WSIMOD also facilitates the representation of boundary conditions within urban systems. For  
503 instance, it allows for the inclusion of abstractions, which are critical in understanding in-river impacts of  
504 wastewater effluent because of their influence on dilution during low flows (Dobson and Mijic 2020).  
505 Besides achieving accurate simulations, we further propose that a modelled representation of boundary conditions  
506 is essential towards assessment of the system under future scenarios. For example, by treating the upstream river  
507 as a model rather than as a fixed boundary condition, the in-river impact of the wastewater system for future  
508 scenarios can be quantified, which would not be possible if the changes to upstream river behaviour under the  
509 future scenario were fixed (Mijic et al. 2022).

#### 510 **4.1.3 Integration to evaluate intervention impacts at whole-water system scale**

511 The OOP deployed in WSIMOD enables flexibility to incorporate a variety of different physical and management  
512 interventions to the water cycle. As demonstrated in Liu et al., (2023a), the included WSIMOD nodes may be  
513 customised both in terms of their parameters and in terms of their physical processes to represent a range of Nature  
514 Based Solutions (NBS) such as flood plains, runoff attenuation features, regenerative farming, urban green space,  
515 and urban wetlands. Because all WSIMOD nodes can communicate with all other nodes, this is a reasonably  
516 straightforward exercise as new rules for interactions with other model components do not need to be redefined  
517 when additional components are added to the model.

518 Due to the flexibility and coverage provided by WSIMOD models, it is often straightforward to include decision-  
519 making, policy constraints, and operational rules (Dobson and Mijic 2020). The high-level control over  
520 orchestration provides an ideal place to represent these stakeholder actions that require information on a multitude  
521 of states across the wider system. For example, the amount available to be abstracted from the River Thames, UK,  
522 is based off both reservoir levels and river flow, meaning no specific node has access to all of the information  
523 required to determine abstraction amount (Environment Agency 1991). Rather than creating a complicated  
524 interaction between London reservoirs and the River Thames, a modeller may more simply inspect the state of  
525 these two systems in the orchestration to dynamically set the abstraction capacity, filling the role of an operator  
526 (Dobson and Mijic 2020). Thus, the nationally important question of modelling changes to abstraction policies  
527 may easily be implemented in WSIMOD, and the impacts of these, both on water supply and in-river water quality,  
528 may be quantified (Mijic et al. 2022).

529 The parsimonious methods selected to represent physical components in WSIMOD ensure that simulations are  
530 computationally efficient. Further to this, if computational speed is of critical interest, then we have demonstrated  
531 that varying timestep and spatial resolution is possible while still achieving accurate simulations (Dobson,  
532 Watson-Hill, et al. 2022). The result is an integrated water cycle model that can be used for purposes that are not  
533 typically computationally tractable. Dobson et al., (2022), demonstrate exploration of uncertainty in sewer model  
534 parameters (roughness and runoff coefficient) is possible with WSIMOD, this study is the first of its kind because  
535 pipe network models are typically too computationally expensive to perform the numerous simulations required  
536 for uncertainty analysis. Meanwhile, Liu et al., (2023a), demonstrate that regional portfolios of 5-year catchment-

537 scale nature-based solutions can be created by applying optimisation to a WSIMOD model spanning 32  
 538 hydrological catchments.

#### 539 4.1.4 Integration to align simulations with systems level outcomes

540 A key benefit we have found in applications of WSIMOD is that, due to the breadth of systems that are represented,  
 541 performance metrics that are relevant to different stakeholders can usually be included in the model in a physically  
 542 based way. The most obvious example, which we have drawn on throughout this paper, is the ability to place in-  
 543 river impacts central to decision making. This enables better alignment with policy goals, such as the Water  
 544 Framework Directive chemical water quality classifications, which are defined based on in-river average pollutant  
 545 concentration (Environment Agency 2020b). However, a variety of other metrics can be conjunctively included  
 546 such as the reliability of water supply. For example, Dobson & Mijic (2020) examine both the water quality and  
 547 water supply benefits of a variety of water cycle interventions (leakage reduction, rainwater harvesting, etc.). Such  
 548 an approach is particularly beneficial to stakeholder engagement because non-water facing stakeholders can still  
 549 be represented and understand their interactions with the water cycle and system-wide goals. For example, Puchol-  
 550 Salort et al., (2022), demonstrate how developers can measure the impacts of new developments on flooding,  
 551 water quality and water supply. Furthermore, this study demonstrates how integrated modelling can enable them  
 552 to quantify how to ensure developments achieve a net-zero impact on these metrics, interestingly revealing that  
 553 retrofitting households outside of the development area is typically required to offset changes.

#### 554 4.2 Data to support water systems integration

555 A key challenge to the application of an integrated approach to water systems modelling is the difficulty of setting  
 556 up models and associated availability of data. In WSIMOD, we provide parsimonious but physically based  
 557 representations to ensure parameterisation is possible with widely available data and we propose model evaluation  
 558 with observed in-river flow and water quality. The catalogue of documented nodes in WSIMOD (Dobson et al.  
 559 2023b) presents data requirements to help users understand how feasible it will be to apply the approach in their  
 560 study area. The data requirements will be entirely dependent on what nodes and at what resolution a model user  
 561 chooses to represent. However, we provide a broad overview of data requirements for a generic catchment scale  
 562 WSIMOD case study in Table 2, which may help to develop future automatic model setup at national and global  
 563 scale to facilitate applicability.

564 **Table 2: List of datasets typically required in catchment-scale WSIMOD applications with references for data sources**  
 565 **or further information.**

Category	WSIMOD input	Availability	Further information (scale)
Climate	Evapotranspiration	Global datasets available	(Khan et al. 2018), (global)
	Temperature	Global datasets available	(Morice et al. 2021) (global)
	Precipitation	Global datasets available	(Sun et al. 2018) (global)
Rural	Hydrological catchment outlines	Global datasets available	(Lin et al. 2019a) (global)
	Hydrological connectivity	Global datasets available	(Lin et al. 2019b) (global)
	Hydrological parameters	Global datasets available for some hydrological models	(Zhang and Schaap 2018) (global)

	Crop surfaces	Global datasets available	(Thenkabail et al. 2016) (global)
	Crop properties	Lookup tables available	(Allen et al. 1998) (-)
	Pollutants/nutrients	National datasets may be available with high uncertainty	(Liu et al. 2022) (UK)
Urban	Population	Global datasets available	(Leyk et al. 2019) (global)
	Garden area	National datasets may be available, otherwise rule of thumb may be acceptable	Office for National Statistics (2021) (UK)
	Wastewater treatment plants	European dataset available through Urban Wastewater Treatment Directive	European Commission (2016) (Europe)
	Foul catchments	National datasets may be available, otherwise contacting wastewater companies required	(Hoffmann et al. 2022) (UK)
Water use	Irrigation water use	Global datasets available	(Thenkabail et al. 2009) (global)
	Water resources system	Not typically available, contacting - water supply companies required	
Evaluation	Flow observations	National datasets typically available	(Fry 2010) (UK)
	Water quality observations (river and WWTW)	National datasets may be available	(Environment Agency 2020a) (England)
	Reservoir levels	Not typically available, contacting - water supply companies required	

### 566 4.3 Future work and research direction

567 One of the main concerns that other modellers have expressed regarding a WSIMOD-like approach is the level of  
568 detail with which components are represented. While hydrologists and agricultural modellers are often  
569 comfortable with parsimony and aggregation in their catchment modelling, most other parts of the water cycle  
570 tend towards more complexity in their representations. This desire for complexity is likely due to the detailed  
571 application context in which different models have been developed. For instance, designing a new process in a  
572 wastewater treatment plant inevitably requires a highly detailed model (Hreiz, Latifi, and Roche 2015). However,  
573 questions are being raised in many fields with a tradition of complex modelling about the need for such  
574 complexity. Models of in-river phosphorus (Jackson-Blake et al. 2017), urban flooding (Li and Willems 2020),  
575 and sewer flow (Dobson, Watson-Hill, et al. 2022; Thrysoe, Arnbjerg-Nielsen, and Borup 2019) have shown that  
576 good results can be achieved with simpler approaches. While practical modellers typically question what level of  
577 complexity is necessary to answer their questions, scientific modellers examine the impacts of assumptions to  
578 build evidence around whether they are suitable and under which circumstances. WSIMOD prioritises integration  
579 of the whole-water cycle, which is enabled by reduced-complexity modelling of the system components. In the  
580 examples provided, we demonstrate that sacrificing complexity in terms of detail should be viewed as an  
581 opportunity to better accommodate and contextualise components in the wider water cycle, as well as highlighting  
582 the importance of interactions between components.

583 We see the WSIMOD platform as an ideal opportunity for the environmental modelling community to implement  
584 and compare (or benchmark) different modelling assumptions and examine water cycle impacts. We plan to  
585 continue developing the representation of different components and testing whether more complex representations  
586 can improve simulations. Our focus will be on representations of a wider range of NBS, treatment plants, and  
587 urban sewer network hydraulic structures. Furthermore, we believe that complementing WSIMOD with machine

588 learning representations of components that are too complex to be captured in a physical way can be a promising  
589 approach, thus implementing a "surrogate" strategy (Razavi et al. 2022; Razavi, Tolson, and Burn 2012).  
590 A key opportunity for improving the accessibility of WSIMOD will be in the development of a graphical user  
591 interface (GUI). The current implementation as a Python package makes the software well suited to customisable  
592 and flexible simulations for programmers, but inaccessible to a wide range of potential users. We see a variety of  
593 different approaches towards greater interactivity and visualisation that are not mutually exclusive. A "virtual  
594 decision room" approach may provide an ideal environment for non-technical stakeholders to explore simulation  
595 results and to highlight integrated system-wide impacts (Schouten, van den Hooff, and Feldberg 2016).  
596 Meanwhile, incorporation into GIS-based frameworks such as 3DNet (Todorović et al. 2019) or Google Earth  
597 Engine (Gorelick et al. 2017) would enable more seamless incorporation of pre-processing and provide a suite of  
598 streamlined tools to help users create, edit, and run WSIMOD models.

## 599 **5 Conclusion**

600 We have presented the theoretical underpinning of WSIMOD, which is an open-source software for simulating a  
601 range of urban and rural processes and operations in the integrated water cycle. WSIMOD represents different  
602 components of the water cycle as nodes that are connected by arcs. The nodes that we have discussed throughout  
603 the paper are parsimonious implementations that are conducive towards easy parameterisation and setup. Arcs  
604 convey interactions between nodes that fall under four key categories: pushes (a node has water to go somewhere),  
605 pulls (a node needs water from somewhere), requests (interaction represents flux of water and/or pollutants), and  
606 checks (interaction does not represent flux). This integration framework allows all nodes to communicate with  
607 each other, thus facilitating a flexible method that can accommodate a wide variety of water systems. Because  
608 this approach uses object-oriented programming, WSIMOD enables customisation to capture unconventional  
609 behaviours and implementation of a wide variety of physical and management interventions.

610 In summary, our early case studies show WSIMOD to be a useful and versatile tool for water systems modelling.  
611 We hope to have persuaded other modellers of the importance of an integrated approach and believe the design  
612 philosophy behind WSIMOD can serve as a helpful starting point for understanding integration in their respective  
613 contexts.

## 614 **6 Competing interests**

615 The contact author has declared that none of the authors has any competing interests.

## 616 **7 Acknowledgements**

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621 improved the paper.

622

## 623 **8 Code availability**

624 WSIMOD is provided open-source under the terms of the BSD-3-Clause license. The code can be accessed at  
625 <https://github.com/imperialcollegelondon/wsi> (last access: 2024-03-26), and documentation at  
626 <https://imperialcollegelondon.github.io/wsi/> (last access: 2024-03-26), with further technical details in Appendix  
627 1. The code has been tested up to Python 3.10 and requires minimal dependencies (see website).

## 628 **9 Author contributions**

629 BD and LL created and tested all model code and documentation. All authors were involved in theoretical  
630 development and writing.

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840

## 841 **11 Appendix 1 – technical details in WSIMOD**

### 842 **11.1 A1.1 Tank object to generalise water stores**

843 A particularly important concept in developing new and understanding existing nodes is the Tank object. The  
844 concept of a water store is so common in water systems that a generic Tank object is provided. Tanks streamline  
845 a variety of uses for stores and have a range of ‘children’ that implement travel time (A1.2) and pollutant decay  
846 (A1.3). A node that represents a water store should be a subclass of the Storage class (see Figure 2), which itself  
847 is a generic node wrapper for the Tank object. The simplest case of a tank would be a water supply reservoir,  
848 demonstrated for WSIMOD at a lumped London scale in (Dobson and Mijic 2020). However, many nodes use  
849 stores but in an auxiliary fashion, for example WWTWs have temporary storage tanks while FWTWs have service  
850 reservoir tanks.

851 **11.2 A1.2 Non-flux pollutant changes**

852 By default, everything tracked in a VQIP follows mass balance. However, in the water cycle, many pollutants  
853 undergo transformations due to biological, physical, or chemical processes, and thus preservation of mass may be  
854 insufficient to simulate them. WSIMOD represents the nitrogen/phosphorus cycles in soil (see documentation of  
855 Land nodes) and denitrification/mineralisation/production/macrophyte uptake in rivers (see documentation of  
856 River nodes), based on the equations from (Lindström et al. 2010; Liu et al. 2022).

857 While the transformations that act on chemicals in soils and rivers are well studied in the literature, there are  
858 difficulties in conceptualising bio-chemical that take place in groundwater and sewers, despite agreement that  
859 these are chemically active (Almeida, Butler, and Friedler 1999). As a result, WSIMOD provides a generic two-  
860 parameter method to implement temperature sensitive chemical decay, given by:

	$M_t = M_{t-1}(1 - cd^{T_t - T_{ref}})$	(1)
--	---	-----

861 , where M is the mass of a chemical in each timestep, c is a parameter that determines non-temperature sensitive  
862 decay, d is a parameter that determines temperature sensitive decay, T is the temperature with a reference  
863 temperature ( $T_{ref}$ ) assumed to be 20C. We do not intend that equation (1) can be a substitute for well-researched  
864 and verified process representations, however in our experience using WSIMOD it is an easy and useful option  
865 to improve water quality representations.

866 Wastewater and freshwater treatment processes are well-studied fields. However, simulation models of these  
867 systems require detailed information describing the different treatment technologies and processes that are present  
868 in a specific plant. While we plan to include these types of models in WSIMOD in the future, we have opted to  
869 take a parsimonious approach to treatment modelling under the assumption that most users will not have detailed  
870 information about the plants they model. This approach assumes that the plant performs a single operation, based  
871 on equation (1), to transform influent, which is then split into three streams of effluent, liquor, and solids.  
872 Depending on whether freshwater or wastewater treatment, these streams go to different places, see documentation  
873 of WTW for further details.

874 **11.3 A1.3 Travel time of water**

875 Arcs are the key model element to implement travel time of water. Two arc subclasses that provide alternate  
876 methods to implement travel time are provided in WSIMOD. The first, more simple approach, formulates the  
877 travel time of the arc as a dictionary object where each key is the number of timesteps remaining; when water is  
878 sent along the arc, it is combined with the any existing water for the key that matches the specified travel time.  
879 These travel times are updated at the end of each timestep. This method is computationally efficient because the  
880 number of operations each timestep is limited by the maximum number of timesteps the arc takes to traverse.  
881 However, this approach cannot represent a dynamic flow capacity, as is the case in, for example, sewer networks,  
882 where hydraulic head governs flow. Thus WSIMOD also contains a less computationally efficient arc to  
883 accommodate this behaviour, described and demonstrated in (Dobson, Watson-Hill, et al. 2022). Arcs can also  
884 implement pollutant changes associated with decay over this travel time, using equation (1).