

1 Scenarios of intermittent *E. coli* contamination from sewer overflows to 2 shellfish growing waters: the Dart Estuary case study

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17 Abstract

18 Sewage overflows (SOs) and Combined Sewer Overflows (CSOs) significantly contribute to the
19 bacterial contamination of coastal waters, which is of especial concern for aquaculture, a growing
20 industry worldwide. Hydrodynamic and water quality models were used to investigate impacts of
21 CSO discharge frequency and duration, river discharge and tides on *Escherichia coli* levels at
22 shellfish farming sites in the Dart Estuary (UK), being the employed methodology generally
23 applicable. High *E. coli* contamination occurred during neap tides and high river discharges due to
24 higher retention and lower bacterial decay. Synchronicity of CSO spills affected the duration of the
25 pollution episodes rather than peak concentrations, more influenced by discharges of the neighbouring
26 CSOs. During peak discharges, *E. coli* concentrations could be 10 times higher than during average
27 flows. CSO spills were more frequent when rainfall was >20 mm. Model outputs combined with
28 rainfall forecasts can indicate microbiological contamination risk in the aquaculture sites.

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30 Highlights:

- 31 • A 3D-hydrodynamic and *E. coli* dispersal model was developed for the Dart Estuary.
- 32 • *E. coli* inputs from sewer overflows peak during neap tides and high river discharge.
- 33 • *E. coli* concentration is highly dependent on the decay model.
- 34 • Sewer overflow spills in the Dart are frequent when rainfall exceeds 20 mm.
- 35 • Model can be used to predict *E. coli* contamination at shellfish farming sites.

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Keywords

Sewage contamination, Dart Estuary (UK), impact assessment, TELEMAC, faecal indicator organisms, shellfish health.

1. Introduction

Sewage contamination is one of the most important issues affecting coastal water quality and sewer overflows (SOs), i.e. overflows from combined (foul and surface water) sewers and from storm water holding tanks at sewage treatment works, feature as the second most important source affecting the quality of surface waters in Europe (European Environment Agency, 2018). To provide clarity, in this study we will use the term Combined Sewer Overflows (CSO), although both types exist. In the UK, a substantial proportion of serious pollution incidents to water are caused by CSOs (Environment Agency, 2016) and future improvements to these discharges are considered key to meet microbiological standards set out by water quality legislation (Defra, 2012). To achieve this, water quality managers require tools to characterise discharge frequency and volumes from CSOs, and the fate and transport of contaminants in receiving waters (Environment Agency, 2018). Modelling has been an integral part of many coastal pollution reduction programmes for characterisation of both quantitative and qualitative aspects of CSO discharge effects. Models used for these purposes usually comprise a hydrodynamic component to represent the advection and dispersion of the discharges and a water quality component to represent the quality of the discharge and subsequent contaminant fate (National Research Council, 2009). In general, the former is easier to calibrate and verify than the latter because contaminant fate and transport are more complex to simulate (National Research Council, 2009) and data are not readily available. This is particularly the case of contaminants such as *E. coli*, which are highly episodic and can vary by several orders of magnitude in very short time and space increments.

In this paper we focus on the Dart Estuary, located on the south west coast of England. Commercial use of the estuary includes shipping, marine services, fisheries and tourism, the latter being mostly water based (e.g. boating, fishing, canoeing). The cultivation of Pacific oysters (*Crasostrea gigas*) in the Dart Estuary dates to the 1960s and there are several sites currently used for this purpose at Waddeton and Long Wood (Cefas, 2011). UK water and sewerage companies have made substantial investments to improve the performance of wastewater treatment and reduce the number of water pollution incidents in shellfish water protected areas (SWPAs) (Environment Agency, 2018; Environment Agency & Natural England, 2017). In the Dart catchment, the largest sewage treatment works (STW), Totnes STW and Dartmouth STW, have been upgraded to include full treatment and ultraviolet (UV) disinfection or membrane bioreactors. However, despite these improvements, the SWPA in the upper Dart Estuary at Waddeton has not complied with the Guideline (G)

1 microbiological standard of the Shellfish Water Protected Areas Directions (SWPAD) 2016 (formerly
2 prescribed by the Shellfish Waters Directive) since its designation in 1999 (Environment Agency,
3 2015). Furthermore, there have been periodic downgrades in the microbiological classification of
4 shellfish farming sites (Cefas, 2010), indicating that the shellfish have been frequently exposed to
5 microbiological pollution in this part of the estuary. Currently, the SWPA is affected by CSO
6 discharges to the estuary, particularly during prolonged rainfall events (Environment Agency, 2015).
7 Shellfish in the Dart rapidly accumulate peak levels of *E. coli* and maintain these for several days
8 after the rainfall events (Campos et al., 2011). This contamination pattern has been observed in other
9 catchments in England where CSOs contribute a substantial proportion of microbiological
10 contamination to coastal waters during high-flow events (Crowther et al., 2011). Policies introduced
11 to mitigate the effect of CSO discharges include the permit system for shellfish waters, operated by
12 the Environment Agency.

13 To reduce the number of CSO spills to an average of 10 spills/annum and help achieve the
14 microbiological objective for the Dart SWPA by 2027, the Environment Agency (EA) recommended
15 further work to reduce diffuse pollution from agricultural land and resolution of operational issues at
16 STWs to reduce the frequency and duration of CSO spills (Environment Agency, 2015).

17 In this paper, we report results of a three-dimensional (3D) hydrodynamic and water quality model
18 for the Dart Estuary to investigate the impacts of CSOs on the microbiological quality of the SWPA at
19 Waddeton. The focus of this modelling study is *E. coli*, which is the bacterial indicator used to classify
20 the extent of microbiological (faecal) contamination in shellfish harvesting areas (European
21 Parliament and Council of the European Union, 2019) and verify compliance with the SWPA
22 Directions G standard (Defra and Natural Resources Wales, 2016).

23 Hydrodynamic models in both 2D and 3D form have been used to simulate estuarine
24 hydrodynamics in many countries for many years. The TELEMAC system in particular, is well suited
25 to this type of study as its unstructured grid and finite element structure allows the best combination
26 of high resolution and computational efficiency. By adopting a baroclinic, 3D version, we aimed to
27 recreate processes of temperature and salinity development and the saltwater stratification and
28 incursion during tidal cycle. These are important features to incorporate salinity and temperature into
29 the computation of *E. coli* concentration in the flow. This is, to our knowledge, the first modelling
30 study of this kind carried out in the Dart Estuary.

31 The model presented here was forced with averaged and peak CSO overflows from 28 discharge
32 points to help us better understand the response of *E. coli* concentrations to variable water flows in the
33 estuary (as determined by river discharges and tides). A preliminary study on the rainfall conditions
34 that induced the CSO spills was also carried out, which constitutes a first step towards a forecasting
35 system of *E. coli* pollution in the Dart Estuary. The model can be used as a tool to provide water
36 resource managers with information to support pollution reduction measures and improve compliance

1 of the shellfish water with microbiological standards and, consequently, to achieve better and more
2 consistent microbiological quality of harvested shellfish.

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4 **2. Material and methods**

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6 *2.1 Study site*

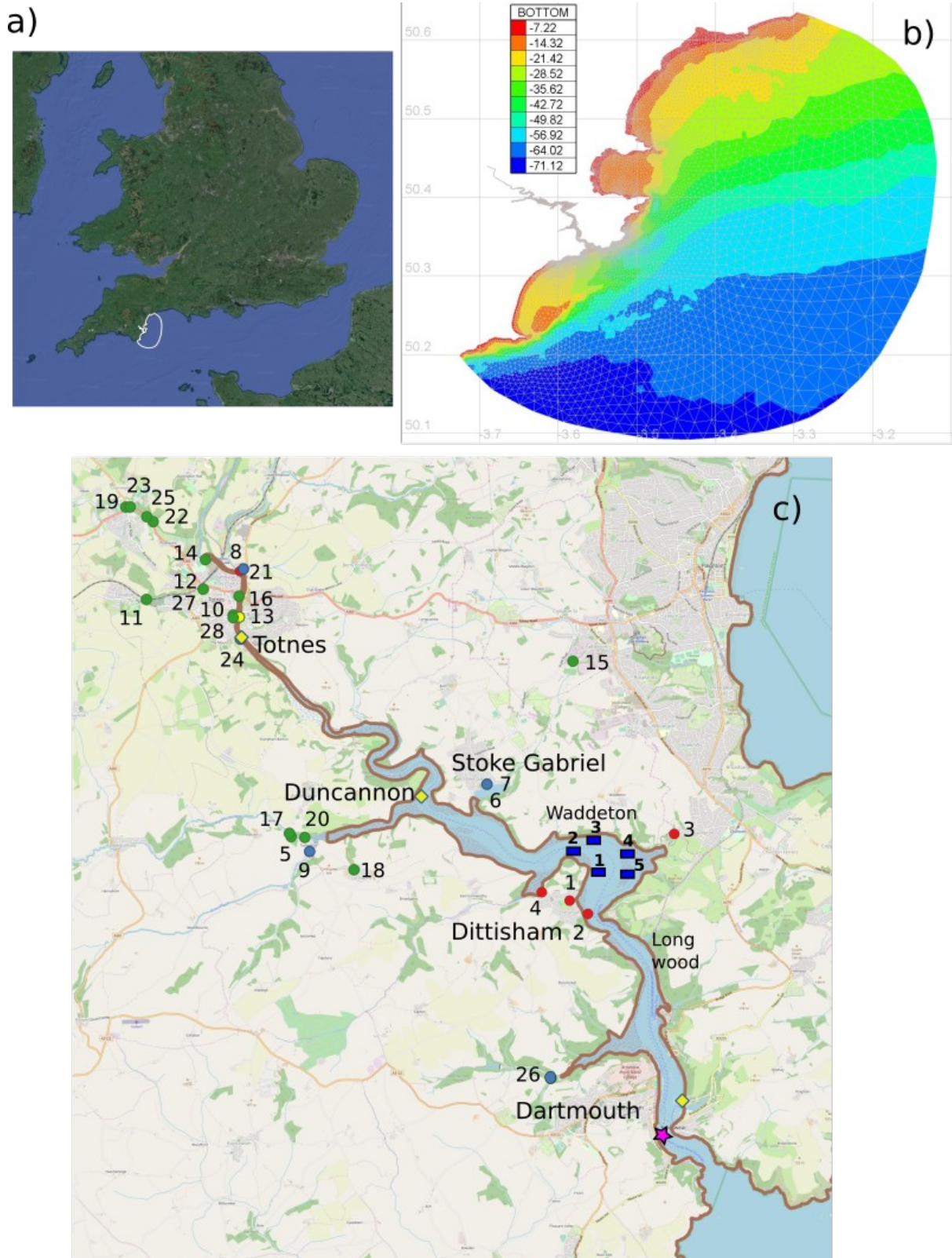
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8 The Dart Estuary and its catchment are located in Devon, on the south west coast of England (Fig.
9 1). Its catchment is an area of 470 km² with moorland drained by steep wooded valleys in the upper
10 catchment and relatively low-lying undulating agricultural land in the lower reaches. The catchment is
11 rural, relatively sparsely populated, with populated centres, Totnes at the head of the estuary, and
12 Dartmouth at its mouth. The estuary is sheltered, branched with small tributaries, and narrows
13 significantly towards its upper reaches. It has a macrotidal regime (5.2 m on spring tides; 1.8 m on
14 neap tides) and is an ebb-dominant, type 3 ria (ABPmer and HR Wallingford, 2007). The upper tidal
15 limit of the estuary is at Totnes, approximately 17 km upstream of Dartmouth. Sandflats, mudflats and
16 a few areas of saltmarsh make up most of the intertidal area. In the lower reaches of the estuary,
17 facilities such as slipways, marinas, moorings and boatyards are common.

18 The main freshwater input to the estuary is the Dart (mean flow=11.4 m³/s measured at Austins
19 Bridge), at the head of the estuary at Totnes (NERC, 2018). Rivers Hems and Harbourne contribute
20 smaller volumes of freshwater. Discharge rates from all these rivers increase rapidly during rainfall,
21 due to low absorption of the hard geology and steep topography of the upper catchment.

22 South West Water (SWW) operates sewerage networks serving five water company STWs within
23 10 km of the tidal limit. The largest discharge volumes are from Totnes STW [(dry weather flow
24 (DWF)=3,967 m³/day, number 8 in Fig. 1) and Dartmouth STW (DWF=4,644 m³/day, number 26 in
25 Fig. 1] (both receive UV disinfection) and from two small STWs (Harbertonford STW, DWF=242
26 m³/day; Ashprington STW, DWF=98 m³/day) (both receive secondary treatment) (Fig. 1). The
27 network is also served by more than 20 CSOs, half of which discharge near Totnes at the head of the
28 estuary and the other half discharge to rivers or directly to the estuary tidal waters, between Dittisham
29 and Stoke Gabriel.

30



1 Fig. 1. Location of the study area (outlined in white) on the south west coast of England (a), model
 2 bathymetry and mesh (b), and Dart Estuary showing the location of the sewer overflows (CSOs) and
 3 shellfish farming sites considered in the study (c). The red circles indicate the CSOs for which spill
 4 frequency and duration data are available. The blue circles indicate the CSOs that were not considered
 5 in this study. The green circles indicate the remaining CSOs. The blue squares indicate shellfish

1 farming areas within the Waddeton shellfish water. The yellow diamonds are the locations of the tidal
2 gauges for which water level data are available and the pink star is a location used to validate the
3 model tidal currents.

4

5 *2.2 Characterisation of storm overflows in the study site*

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7 In England, the EA applies a set of standards to the determination of permit (consent) applications
8 for discharges that impact on shellfish waters. For shellfish waters impacted by multiple CSOs, the
9 EA recommends aggregating spills by frequency and volume so that the combined impact of the
10 aggregated spills does not exceed 10-spills per annum or 3% of the time on average (Environment
11 Agency, 2003). However, spills do occur beyond those permitted by the regulations.

12 Twenty-eight CSOs were identified in the Dart catchment (Fig. 1 and Table 2) for this modelling
13 study. Data on the frequency and duration of sewage spills from these overflows were available from
14 SWW annual spill monitoring reports to the EA (relative to eight CSOs over the period 1 April 2006–
15 31 March 2016, Fig. 2) and sewer network modelling reports (Frischmann, 2001; Hyder, 2012). Time
16 series analysis of data taken from the EA annual reports shows that six CSOs (Totnes STW SO, Stoke
17 Gabriel SPS, Higher Dittisham SPS, Galmpton PS, Ferry Boat Inn PS and Dittisham STW SO) have
18 spilled over longer time periods and/or more frequently than the remaining ones (Fig. 2), with some
19 overlapping among them. For the remaining CSOs, information on the number and duration of spills
20 was obtained from modelling results. In particular, information for the Dartmouth STW was obtained
21 from the Frischmann (2001) report, which briefly describes a model implemented for this CSO
22 applying data for 2001. For the remaining CSOs, Hyder (2012) specifies that the catchment modelling
23 software Infoworks was used to simulate the possible spills into the Dart Estuary from the different
24 catchment areas for summer 2011 (since data were available for model validation in this period) and
25 included a prediction for 2036. The data used to implement these models were: resident population
26 connected to the sewerage catchment, rainfall records in the period 2008–2011 provided by the EA,
27 SWW data obtained through the Supervisory Control and Data Acquisition (SCADA) and the EA
28 Monitoring Certificate Scheme (MCERTS) systems, SWW spill monitoring reports, SWW record
29 drawings and operations information, SWW 2008–2009 STW DWF Base Flow Study, EA defined
30 agglomerations and South Hams District Council Local Plan Review (1995–2011).

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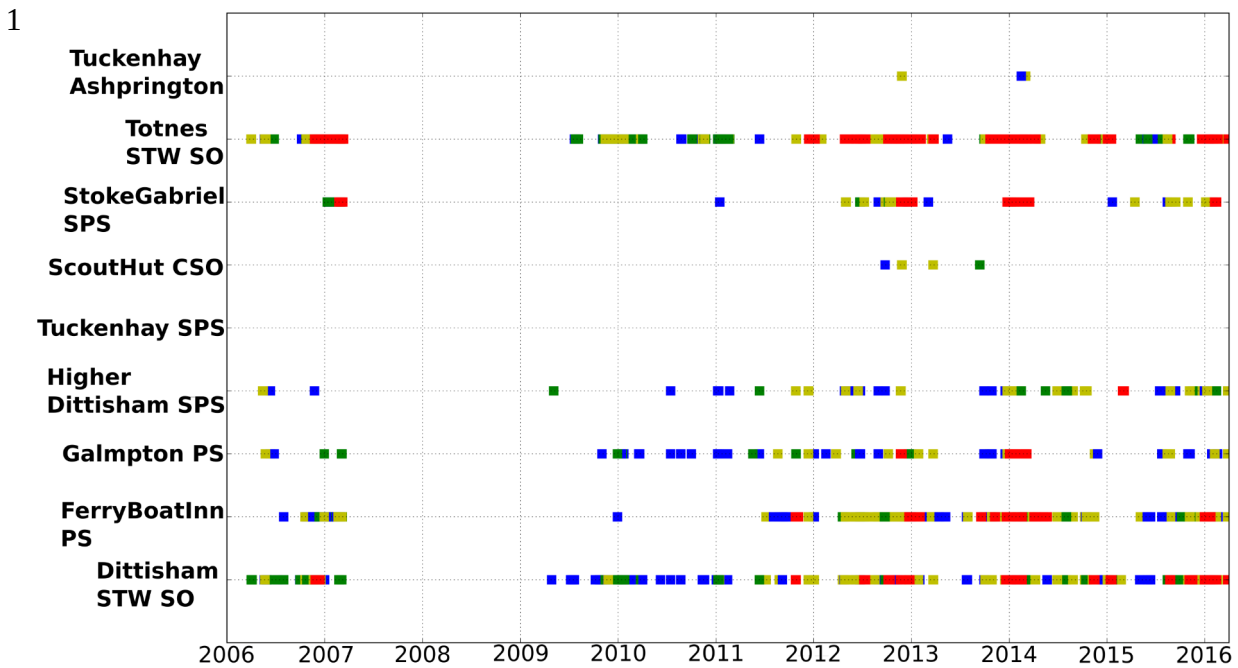


Fig. 2. Sewage spills from eight storm overflows in the Dart Estuary. In blue, spills < 12 h; in green, spills lasting 12–24 h; in yellow, those lasting 24–72 h and in red spills > 72 h.

2.3 Shellfish water regulations and improvements to storm overflows

The UK Government has identified water quality objectives and relevant statutory obligations that water and sewerage companies must observe to meet the requirements of the Water Framework Directive. For SWPAs, the objective is to endeavour to observe a microbiological standard of 300 *E. coli*/100 ml in shellfish flesh for 75% samples in a 12-month period (Defra and Natural Resources Wales, 2016), as determined by the SWPAD. In addition to this standard, the Food Hygiene Regulations (primarily Regulation (EC) No 627/2019) prescribe the harvesting area classification categories for products intended for human consumption. Under these regulations, the class A standard is 230 *E. coli*/100 g in 80% of samples with an upper limit of 700 in the remaining 20% of samples. Shellfish from production sites with *E. coli* concentrations $\leq 4,600/100$ g (in 90% of samples, with no sample exceeding 46,000 *E. coli*/200 g) are class B.

For the purposes of water company investment planning for 2020–2025, the EA’s design standard for intermittent discharges is ≤ 10 significant spills/year as 10-year average. For this, the significant spill volume for the asset or an agglomeration of assets (that impact collectively) needs to be determined, based on a shellfish water receiving water geometric mean of 5 *E. coli* cfu/100 ml. The affordability of improvements will be subject to cost/benefit criteria and ministerial approval. Where the scheme is not cost-beneficial, a possible future target for CSOs for use in planning investment in 2020–2030 is the 10 significant spills criterion against a geometric mean of 110 *E. coli* cfu/100 ml in

1 the SWPA. This equates to meeting the class B standard of Regulation (EC) No 627/2019 in shellfish
2 flesh.

3 In the environmental investment programme proposed by SWW for the period 2020–2025, several
4 CSOs are identified for improvement in relation to the Dart SWPA (see Table 1).

5

6 **Table1.** CSOs identified for improvements by SWW in relation to the Dart SWPA in the period 2020-2025

STW/CSO Infrastructure	Figure 1 Reference
Ashprington STW SSO	5
Cornworthy STW SO	18
Fore Street CSO	10
Lower Collapark CSO	27
Northern Villages PS CSO	21
Quarry Close CSO	11
Scout Hut CSO	6
St Johns Terrace CSO	12
St Katherine's Way CSO	28
St Peter's Quay PS CSO	24
Steamer Quay CSO	13
Stoke Gabriel PSCSO/EO	7
Swallowfields Kevics CSO	14
Tor Park Road PS CSO/EO	15
Totnes STW CSO	8
Totnes Town PS CSO/EO.	16

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8 *2.4 Hydrodynamic model setup*

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10 A three-dimensional hydrodynamic model of the Dart Estuary was built using TELEMAC 3D
11 (v7p2r2; EDF, 2017). The model domain comprises the estuary and the adjacent coastal and offshore
12 waters to allow for a better propagation of the boundary conditions and to investigate the variability of
13 the river plume (Fig. 1b). The domain was discretized by means of an unstructured grid with 12,054
14 nodes and 22,429 elements in the horizontal and 10 equally spaced layers in the vertical. The
15 refinement in the grid varied spatially, with higher resolution inshore and coarser offshore (resolution
16 range=13-3,843 m). The model bathymetry was mostly obtained from UKSeaMap (2010) with some
17 modifications in the upper estuary to properly reproduce tidal propagation. The boundary conditions
18 for the velocities and surface elevations on the offshore open boundary were obtained from the OSU
19 TPXO European Shelf 1/30° regional model (11 tidal constituents: M2, S2, N2, K2, K1, O1, P1, Q1,
20 M4, MS4 and MN4, see <http://volkov.oce.orst.edu/tides/ES.html>).

21 Temperature and salinity were kept constant in space and time along the boundary (12.2 °C and
22 35.1, respectively). Freshwater inputs from the River Dart were accounted for in the model by
23 imposing a time series of river runoff as measured at Austin's Bridge. Water temperature and salinity
24 were set at a constant value of 12.2 °C and 1.0, respectively. It must be noted that, since temperature
25 was defined as a “constant” and no atmospheric forcing was considered in model simulations, water
26 temperature remains constant over time. This approximation is believed not to affect the modelled

1 water circulation in the estuary because the density structure of the water column is mainly driven by
 2 salinity variations, being the contribution of temperature negligible (see Priestley, 1998).

3 In this study, we considered the non-hydrostatic version of the Navier-Stokes equations. The κ - ϵ
 4 turbulence model was selected for the horizontal and vertical dimensions. Wetting and drying were
 5 included in the calculations due to the presence of tidal flats. Advection schemes that ensure
 6 conservative and monotonic behaviour were selected for tracers (scheme 14 in EDF, 2017). The
 7 model was run for a two-month period, starting on 1 December 2015, to capture periods of high and
 8 low river discharge and different tidal phases (Fig. 8). The initial seven days of model simulations
 9 were considered the spin-up period and hence were not used in the analysis.

10

11 2.5 Modelling *E. coli* transport and decay

12

13 The equation used to model the transport and decay of *E. coli* can be expressed as:

$$\frac{\partial EC}{\partial t} + \vec{u} \cdot \vec{\nabla} (EC) = (\vec{K} \cdot \vec{\nabla} (EC)) - k_d EC, \quad (1)$$

14 where EC is the concentration of *E. coli* measured in cfu/100 ml, \vec{u} is the flow velocity, \vec{K} is the
 15 diffusion coefficient and k_d is the decay rate for *E. coli*. The left-hand side of Equation (1) represents
 16 bacterial advection, the first term on the right-hand side shows the diffusion and the last term accounts
 17 for the exponential decay of *E. coli*.

18 The die-off of faecal bacteria is often represented by a T_{90} parameter, which is defined as the time
 19 for bacterial concentration to reduce by 90%. T_{90} is related to the decay rate k_d by the following
 20 equation:

$$k_d = \frac{2.3}{T_{90}}. \quad (2)$$

21 The decay rate of *E. coli* is a function of several environmental parameters such as temperature,
 22 salinity, solar irradiation or degree of sewage mixing (see, for instance Campos et al., 2013; Carneiro et
 23 al., 2018 or Alkan et al., 1995). Several models have been proposed to quantify either k_d or T_{90} as a
 24 function of some of these variables (see [Chan et al., 2015; Mancini, 1978] or the compilation of
 25 different models made by Feitosa et al., 2013), all of them adapted to the local conditions by means of
 26 certain parameters. An example of the application of this kind of model in UK waters is found in Gao
 27 et al. (2015). A simpler approach in which different constant decay rates for day and night have been
 28 considered is published in Abu-Bakar et al. (2017) and Huang et al. (2017).

29 In this modelling study, we used two approaches to calculate *E. coli* decay rates:

- 30 1. An “explicitly variable” decay rate (EVDR) dependent on temperature (T, in degrees Celsius),
 31 salinity (S) and solar irradiation (I, in Watts/m²), based on Mancini, (1978). Namely,

$$k_d(T, S, I, z) = [0.8 + 0.017 S] \times 1.07^{(T-20)} + 0.086 \times I \times e^{-k_z z}, \quad (3)$$

1 where z is the water depth and k_z is the light extinction coefficient, which in TELEMAC 3D is
2 calculated as $1.7/z_{SD}$, with z_{SD} the Secchi disk depth, fixed at a value of 0.9 m for this study.

3 Note that Mancini's equation is based on data from New York estuary data. Although Gao et
4 al. (2015) equation is based on UK estuaries, we did not consider it in this paper because not
5 all the values of the equation coefficients were included in the paper.

- 6 2. An "implicitly variable" decay rate (IVDR) considering the values for estuarine waters of
7 Huang et al. (2017): one constant value of $T_{90}=8.6$ h for day time and another one of 30.64 h
8 for night time. These values are within those reported by Abu-Bakar et al. (2017) in a
9 different UK estuary (T_{90} ranged between 2.5 and 20 h for day time and between 30 and 60 h
10 for night time).

11 Both approaches were implemented in the TELEMAC subroutine SOURCE_TRAC. In the first
12 case (EVDR), the waqtel process 5 (atmosphere-water exchange) was activated, and the solar
13 irradiation (I) in Equation (3) was calculated by means of the TELEMAC subroutine SOLRAD. The
14 inputs provided to SOLRAD were the latitude, longitude and cloud coverage (obtained from ECMWF
15 ERA-interim - www.ecmwf.int/en/research/climate-reanalysis/era-interim - in the geographical centre
16 of the computational domain). The date and time of the day is automatically read during the model
17 calculations.

18 19 *2.6 Data and assumptions on sewer overflows considered in the model*

20
21 Overflow discharge volumes and average spill duration for each CSO were used as input into the
22 model by combining the information in Section 2.2 (Table 2). The annual average discharge volumes
23 (m^3/day) were calculated using data reported by Frischmann (2001) (for Dartmouth STW) and Hyder
24 (2012) (the remaining CSOs, based on the 2011 modelling results). The peak discharges (m^3/day)
25 were provided by SWW. The average spill duration was extracted from the SWW annual reports,
26 when available (CSOs number 1-9 in Table 2) and from the Frischmann (2001) (for Dartmouth STW)
27 and Hyder (2012) reports for the rest of the CSOs.

28 Information on average spill duration and/or average flow was not available for six CSOs
29 (identified with an asterisk in Table 1). Of these, Stoke Gabriel SPS, St. Peter's Quay, Ashprington
30 N°1 and Northern Villages were removed from the simulations; the first three because they do not
31 spill during annual events and the last one for being an emergency overflow, meaning that it should
32 only operate during pump or rising main failure. For Tuckenhay and Ashprington SPS N°2 and
33 Malsters SPS, missing information was assigned the median of the data for the remaining CSOs used
34 in the model simulations (indicated with two asterisks in Table 2).

35 Monitoring of *E. coli* in effluents from CSOs is not required under the EA discharge consenting
36 policy (Environment Agency, 2003). Therefore, we used as input concentrations into the model the

1 reference *E. coli* concentrations for different treatment levels and individual types of sewage-related
 2 effluents under different flow conditions reported by Kay et al. (2008), which are the result of a study
 3 undertaken across the UK in 15 study areas over an 11 year period (1995-2005) at 205 river/stream
 4 monitoring points. Totnes STW-SO and Dittisham STW-SO (stored settled sewage) were assigned a
 5 *E. coli* concentration of 8×10^5 cfu/100 ml, while the remaining CSOs were assigned a concentration
 6 of 2.5×10^6 cfu/100 ml (storm sewage overflows). Effluent from Dartmouth STW is UV disinfected
 7 and has a reported mean *E. coli* concentration of 95 cfu/100 ml, which is several orders of magnitude
 8 lower than those in CSO discharges. For this reason, this discharge was excluded from the modelling.

9

10 **Table2.** Summary of discharge information for twenty-eight sewer overflows considered in the
 11 modelling study.

Name of storm overflow	Figure 1 reference number	Average spill duration (hours)/number of spills	Average discharge volume (m ³ /day)	Peak discharge (m ³ /day)	Assigned <i>E. coli</i> concentration (cfu/100 ml)
Dittisham STW SO	1	21.3 / 326	568	4,838.4	8×10^5
Ferry Boat Inn PS SPS Galampton PS	2	23.8 / 258	77.7	4,147.2	2.5×10^6
SPS Higher	3	14.7 / 179	785.9	5,529.6	2.5×10^6
Dittisham SPP Tuckenhay and Ashprington	4	7.8/ 190	87	2,505.6	2.5×10^6
SPS No2*	5	0.2 / 4	301.6**		2.5×10^6
Scout Hut CSO	6	7.5 / 12	858	18,144	2.5×10^6
Stoke Gabriel SPS*	7	47/ 69			2.5×10^6
Totnes STW SO	8	48.8 / 271	201	8,812.8	8×10^5
Tuckenhay and Ashprington	9	4.5 / 16			2.5×10^6
SPS No 1*	10	7.1	217	26,265.6	2.5×10^6
Fore Street CSO	11	2.4	226	11,923.2	2.5×10^6
Quarry close CSO	12	0.5	358	5,270.4	2.5×10^6
St. Johns Terrace CSO	13	3	1,344.9	41,904	2.5×10^6
Steamer Quay CSO	14	5.3	253.3	12,873.6	2.5×10^6
Swallowfields Kevics CSO	15	1	640	1,209.6	2.5×10^6
Tor Park Road PS/CSO/EO					

Totnes Town PS/CSO/EO Ashprington	16	10	1,581.9		2.5 x 10 ⁶
STW SSO Cornworthy	17	17	34	2,160	2.5 x 10 ⁶
STW SO Dartington	18	5.2	22.4	1,900.8	2.5 x 10 ⁶
School 2 CSO Malsters SPS*	19	4.8	301.6	18,230.4	2.5 x 10 ⁶
Northern Villages PS/EO*	20	5.3**	301.6**		2.5 x 10 ⁶
Queens Arms CSO	21				2.5 x 10 ⁶
Shinner's Bridge SCO	22	6.8	107	604.8	2.5 x 10 ⁶
St. Peter's Quay PS/CSO*	23	5.1	97	1,296	2.5 x 10 ⁶
Textile Mill CSO	24				2.5 x 10 ⁶
Dartmouth STW Lower Collar	25	9	303	5,011.2	2.5 x 10 ⁶
Park St. Katherine's	26	1.1	540,072		9.5 x 10 ¹
Way	27	4	428.5	26,784	2.5 x 10 ⁶
	28	2.8	641.7	41,212.8	2.5 x 10 ⁶

1 * Some data were not available for the CSO.

2 ** Value assumed to be equal to the median of all other CSOs.

3 NB. Sewer overflows with missing data on average spill durations and/or average discharge volumes
4 were not considered in the model simulations.

5

6 2.7 Model scenarios

7

8 To investigate the impact of CSO discharges on the Dart Estuary, we considered the seven
9 scenarios listed in Table 3. The specific aims of these scenarios were to study:

10

- 11 • The effects of river discharge and tidal phase on the fate and transport of *E. coli* (Scenarios 1–
12 4).
- 13 • The effect of CSO spills from all discharges occurring simultaneously (worst-case) (Scenario
14 2 versus Scenario 5).
- 15 • The difference between average and peak discharge conditions (Scenario 2 versus Scenario
16 6).
- 17 • The effect of *E. coli* decay rate on modelling results (Scenario 2 versus Scenario 7).

1 **Table 3.** Details of the modelling scenarios considered in the study.

Scenario Number	River discharge	Tidal phase	Sewer overflows considered	<i>E. coli</i> decay rate	Simulation start date and time	Simulation length (days)
1	High	Neap	Five quasi-simultaneous spills in Fig. 2 (Totnes_STW_SO, Stoke Gabriel SPS, Higher Dittisham SPS, Galmpton PS, Ferry Boat Inn PS and Dittisham STW SO. Notice that Stoke Gabriel SPS has been removed as explained in Section 2.6)	EVDR	04/01/2016 (13:00)	5
2	Low (Fig. 8(a)–(b))	Neap	spilling at average flow rates Five quasi-simultaneous spills in Fig. 2	EVDR	19/01/2016 (13:00)	5
3	Low	Spring	spilling at average flow rates Five quasi-simultaneous spills in Fig. 2	EVDR	13/01/2016 (09:00)	5
4	High (Fig. 8(a)–(c))	Spring	spilling at average flow rates Five quasi-simultaneous spills in Fig. 2	EVDR	26/01/2016 (07:00)	5
5	Low	Neap	All the CSOs in Table 1 for which data are available, except Dartmouth STW spilling at average flow rates	EVDR	19/01/2016 (13:00)	5
6	Low	Neap	Five quasi-simultaneous spills in Fig. 2 spilling at peak flow rates	EVDR	19/01/2016 (13:00)	5
7	Low	Neap	Five quasi-simultaneous spills in Fig. 2 spilling at average flow rates	IVDR	19/01/2016 (13:00)	5

1 2.8 Data for model validation

2

3 As a general comment, the data available for model validation were scarce in time and space, except
4 for the water levels. Some of the data presented in this section were not directly used for the model
5 validation but are included to illustrate the range of values that are expected for different variables in
6 the Dart Estuary.

7

8 2.8.1 Water levels

9

10 Water levels (m) at Totnes (tidal limit), Duncannon (mid-estuary) and Dartmouth (estuary mouth)
11 were obtained from DartNet (<https://www.valeport.co.uk/InsideValeport/DartNetTides>, see Fig. 1(c)
12 for their location) (Valeport, 2018). Tidal prediction data provided by the UK Hydrographic Office
13 (UKHO) were also obtained from the website. Generic current velocity data published in the peer
14 reviewed literature (Thain et al. 2004) were used for model validation because detailed data were not
15 available (see Section 3.1.2).

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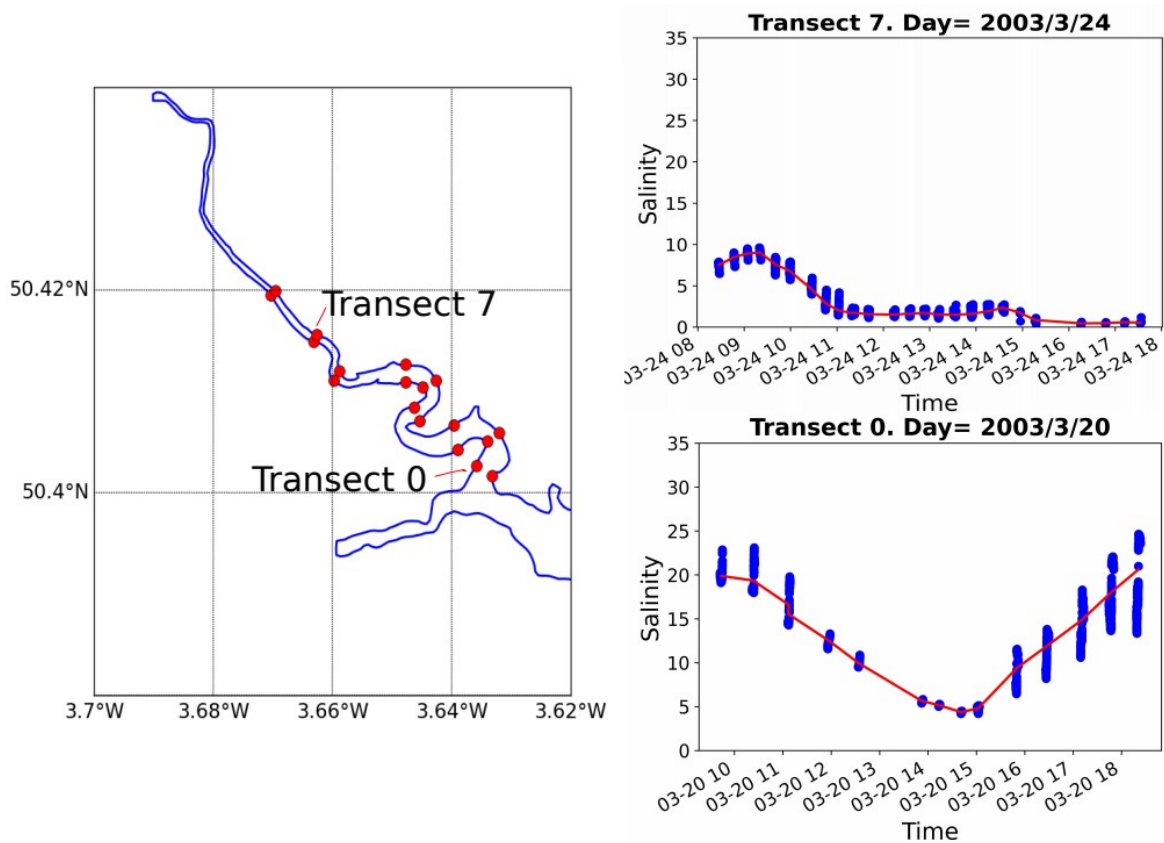
17 2.8.2 Salinity

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19 Salinity data available for eight cross-river transects in the upper estuary for five days (20–25 March
20 2003) during a period of low river discharge ($<10 \text{ m}^3/\text{s}$) were used for model validation. Each transect
21 consisted of several stations (the number depending on the transect width) where salinity was
22 measured using hand-held CTD profilers from a small boat. Each station was sampled at different
23 stages of the tidal cycle at a frequency of 1 s for approximately 3 minutes. Fig. 3 shows salinity
24 results along transect 7 (greater freshwater influence) and 0 (greater marine influence). Salinities were
25 lower and less variable at station 7 than at station 0 indicating the penetration of seawater on the flood
26 and influence of freshwater discharges on the ebb. CTD casts taken at a number of sites in the upper
27 part of the estuary on 19 March 2003 indicate a degree of stratification. It should be reminded that the
28 available salinity data are representative of periods of low river discharge and therefore do not reflect
29 the full range of salinity variation expected in the estuary.

30 Few data exist to describe the variability of water column stratification in the Dart Estuary. The
31 existing data are mostly representative of the estuary mouth. For instance, Priestley (1998) analysed
32 the stratification using data from a telemetered monitoring buoy deployed at approximately 2 km
33 upriver from the estuary mouth over three weeks in March–April 1998 to study water column
34 stratification during a spring tide-neap tide-spring tide cycle. Thain et al. (2004) took CTD casts to
35 measure the water column stratification during spring tide on 13 March 2002 and neap tide on 22
36 March 2002. Both studies have shown that the Dart is a partially mixed estuary with a complete
37 stratification/destratification cycle during the spring/neap tide transition.

1



2 Fig. 3. Measured salinity at sites along transects 0 and 7 on 20 and 24 March 2003. At each site
3 (corresponding to each time cluster) the high resolution raw data (one observation every second for
4 approximately three minutes) measured in the first 40cm of the water column are depicted in order to
5 show the range of variability. The red line shows the average of each time cluster. The Time format in
6 the x axis is mm-dd hh.

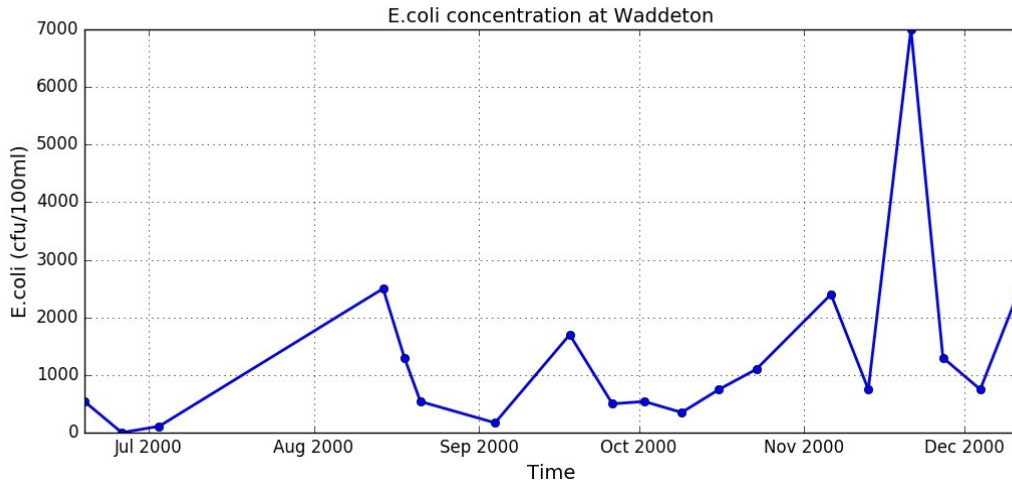
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8 2.8.3 *E. coli* concentrations

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10 *E. coli* concentrations quantified in surface water samples taken weekly at nine sites along the estuary
11 during six months (June–December 2000) were used in this study. Fig. 4 shows that the measured *E.*
12 *coli* concentration at Waddeton, which is close to the shellfish farming areas (Fig. 1(c)) exceeded
13 2,000 cfu/100 ml in August, November and December, and 7,000 cfu/100 ml in December. However,
14 since the time when these samples were taken, the sewerage infrastructure has been upgraded and
15 tertiary treatment installed at Totnes STW and Dartmouth STW. Hence, these values may not be fully
16 representative of the current levels of contamination in the estuary, even in the event of spills and
17 changes in degree of diffuse input of *E. coli* from farming sources.

1



2 Fig. 4. *E. coli* concentrations in surface water samples taken at a site in the Waddeton shellfish water.
3 Data from Allen (2001).

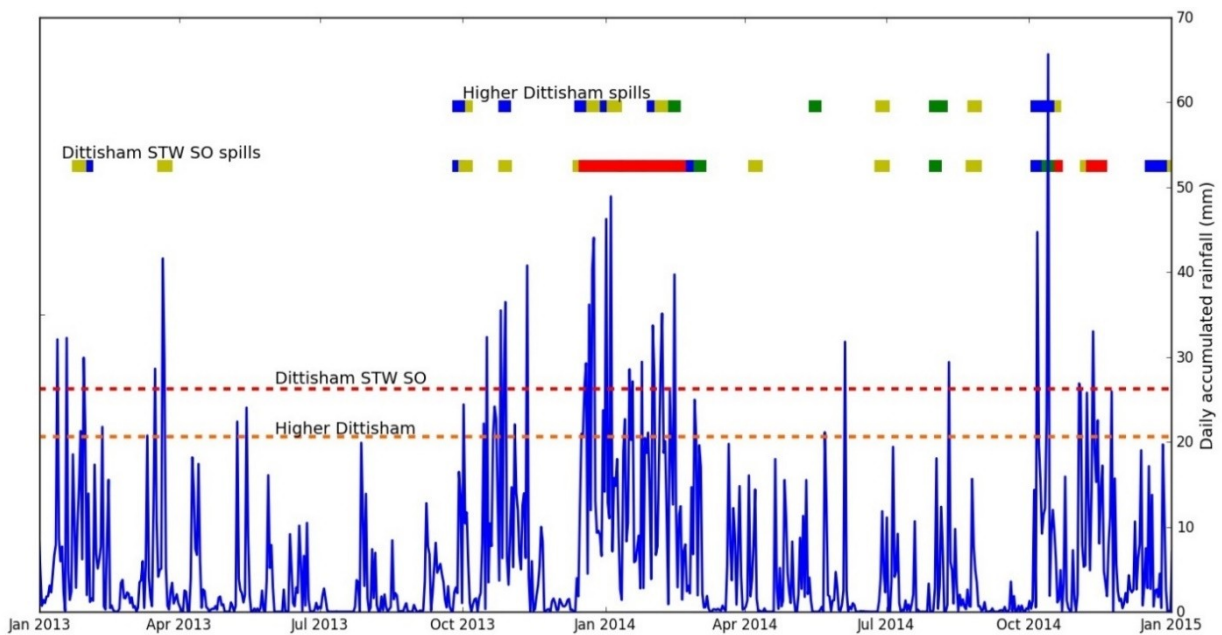
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5 2.8.4 Rainfall

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7 In order to do a preliminary assessment of the correlation between rainfall and CSO spills, daily
8 cumulative rainfall data were obtained from ECMWF ERA-Interim
9 (www.ecmwf.int/en/research/climate-reanalysis/era-interim) at the geographical centre of the
10 computational domain and are displayed in Fig. 5 together with the sewage spills from Higher
11 Dittisham and Dittisham STW SO during the period 2013–2015. In most of the cases, the sewage
12 spills occurred during rainfall events. Higher volume spills generally coincided with periods of high
13 rainfall.

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1 Fig. 5. Daily cumulative rainfall in the Dart Estuary catchment (mm) and sewage spills from Higher
2 Dittisham CSO and Dittisham STW SO, January 2013–December 2014. The dashed lines represent
3 the averaged cumulative rainfall over spill duration. Rainfall data from ECMWF ERA-Interim.
4

4

5 **3. Results**

6

7 *3.1 Validation of the hydrodynamic model*

8

9 *3.1.1 Water levels and current velocities*

10

11 Figs 6(a), 6(c) and 6(e) compare model results with observed data on water levels at the three tidal
12 gauges described in Section 2.8.1 (see Fig. 1 for their locations). The predictions from the UKHO area
13 are also included for reference. Figs 6(b), 6(d) and 6(f) show the relationships between modelled and
14 predicted observations. Model skill was quantified by means of bias, Root Mean Square Error
15 (RMSE) and correlation (Table 4). The results show that the model reasonably reproduced the tidal
16 cycle at all validation sites, although with certain anomalies (i.e. 11 cm mean underestimation at
17 Dartmouth, 11 cm mean underestimation at Duncannon and 1.8 cm mean underestimation at Totnes).
18 Some events such as the high water levels registered simultaneously around the 8th of January 2016 at
19 all tidal gauges were not captured by the model. Some events such as the high water levels registered
20 simultaneously around the 8th of January 2016 at all tidal gauges were not captured by the model. The
21 differences between our model and the observations were very similar to those obtained by the UKHO
22 prediction (Table 4), but variable depending on the location. In both cases (our model and the UKHO
23 prediction), the differences can be attributed to the fact that the observations contain the effect of
24 variable meteorological conditions (i.e. changes in pressure or wind) that affect the water levels,
25 which are not included in the UKHO predictions nor in our model. Differences can also be attributed
26 to reduced propagation of the tidal signal over the extremely variable bathymetry of the Dart Estuary.
27 Although no direct observations of current velocities were available to validate our model, data from
28 the literature indicate that, in the estuary mouth, characteristic velocities on the flood phase of spring
29 and neap tides are ≈ 0.6 m/s and ≈ 0.3 m/s, respectively (Thain et al., 2004). Figs 7(c), and 7(d) show
30 modelled current velocities at a site in the estuary mouth (marked with a star in Fig. 1(c)), which are
31 in close agreement with these values.

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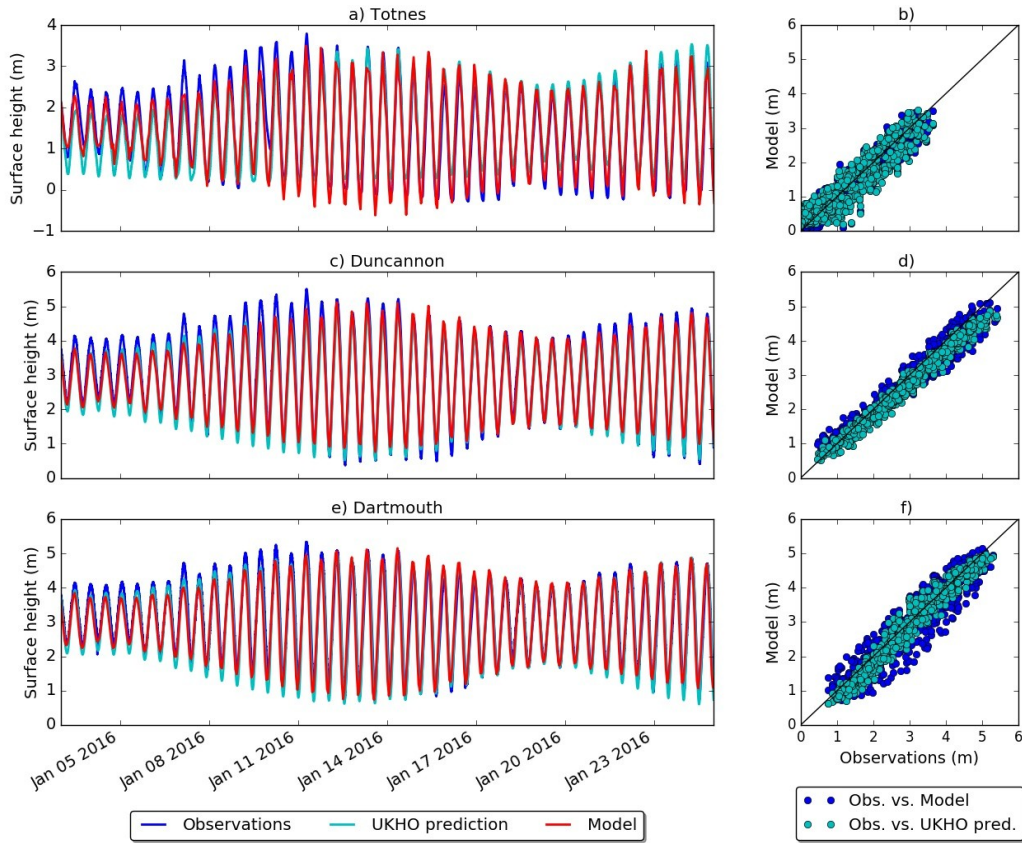
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2 **Table 4.** Statistics of modelled and observed results for three tidal gauges considered in the study.

	Totnes		Duncannon		Dartmouth	
	Model	UKHO	Model	UKHO	Model	UKHO
Bias (cm)	-1.8	0.13	-10.31	-22.38	-11.17	-11.04
RMSE (cm)	30	41.84	32.28	30.24	56.27	26.52
Pearson	0.95	0.91	0.97	0.99	0.91	0.98
Correlation						

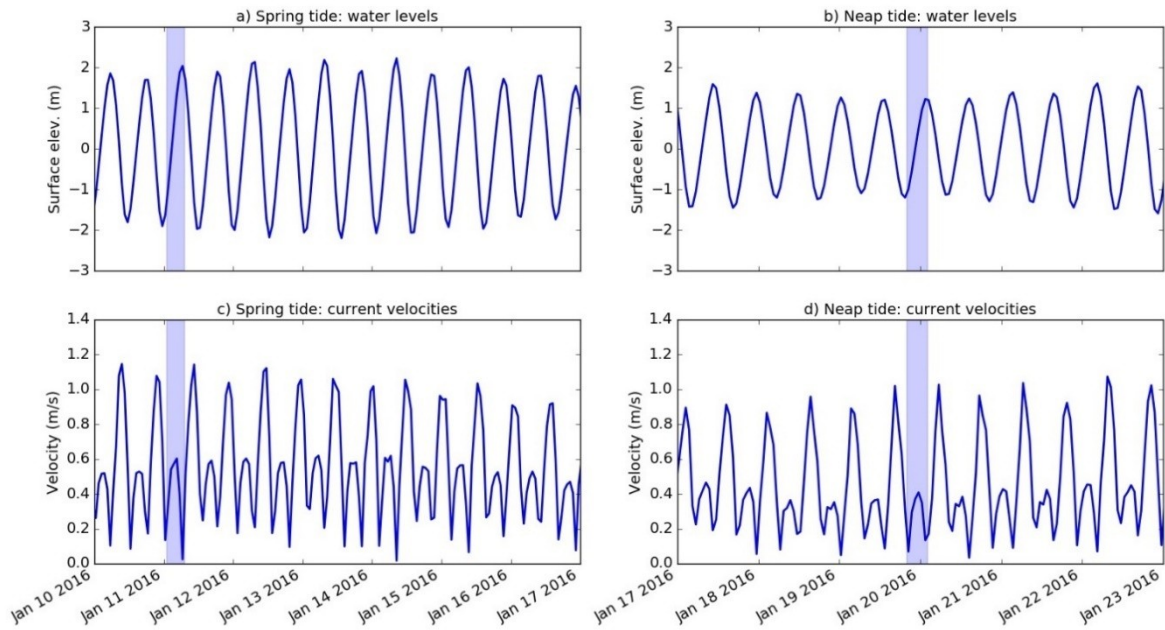
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5 Fig. 6. Time series of modelled and observed water levels at three tidal gauges in the Dart Estuary.

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2 Fig. 7. Modelled current velocities on spring and neap tides in the Dart Estuary mouth, 10–17 January
3 2016.

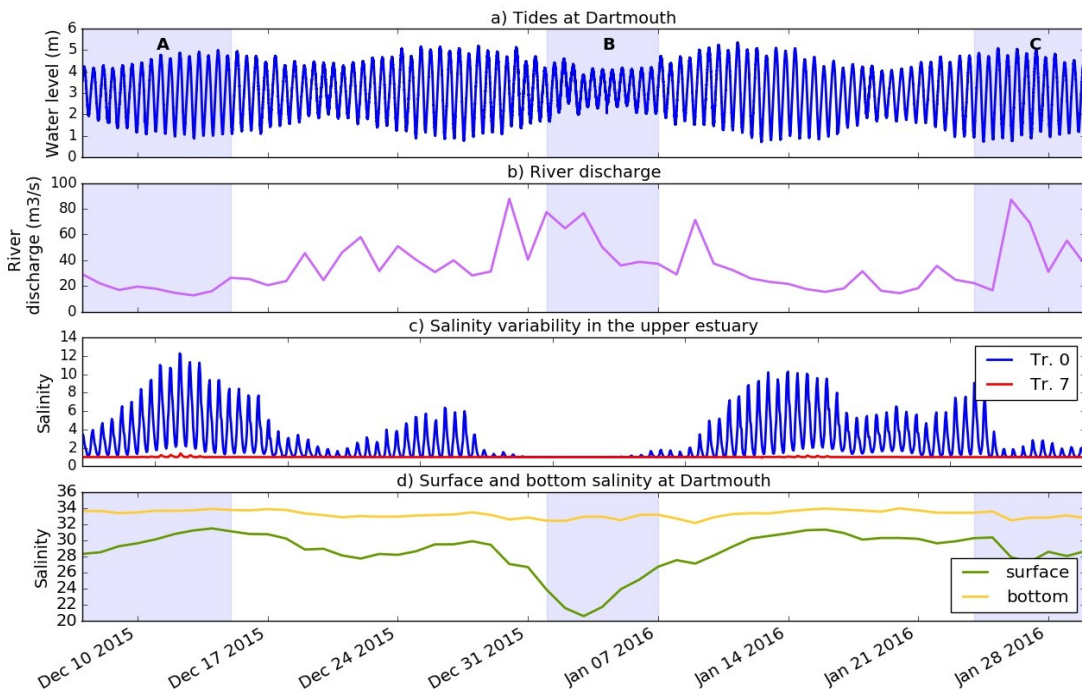
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5 3.1.2 Salinity and water column stratification

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7 Fig. 8 shows the variations of water levels at Dartmouth, river discharge, surface salinity at a site in
8 the middle of transects 0 and 7 (located in the upper estuary, Fig. 3), and the surface and bottom
9 salinities at Dartmouth. The results show the effect of river discharges in modulating the amplitude of
10 salinity levels during flood and ebb tides. Salinities in Fig. 3 reach higher values than those predicted
11 by the model because the river discharge during the modelled period (minimum $\approx 20 \text{ m}^3/\text{s}$, Fig. 8(b))
12 was much higher than that recorded during the period when the salinity data displayed in Fig. 3 were
13 sampled ($\approx 5 \text{ m}^3/\text{s}$).

14 Averaged daily salinities at surface and bottom of the water column are shown in Fig. 8(d). The
15 results show that the tidal regime plays an important role in the stratification/destratification of the
16 water column, as described by Thain et al. (2004) and that this is also determined by the river
17 discharge. This effect is illustrated for three time periods: period (A), when freshwater discharges
18 remain relatively constant and the stratification increases during neap tide; period (B), when high
19 freshwater discharges coincide with neap tide and stratification strengthens to maximum levels; and
20 period (C), when freshwater discharges coinciding with spring tides are sufficiently high to cause
21 stratification of the water column.



2 Fig. 8. Time series of water levels at Dartmouth (a), river flows (b), salinity at transects 0 and 7 in the
 3 upper estuary (c) and salinity at two depths measured at the estuary mouth (d), 10 December 2015–28
 4 January 2016.

5

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7 3.2 Modelling scenarios

8

9 3.2.1 Effect of tides and river discharge on *E. coli* concentrations

10

11 This section reports results on the effect of hydrodynamic conditions (tidal phase and river discharges
 12 according to Scenarios 1–4 in Table 3) on the fate and transport of *E. coli* in the Waddeton SWPA
 13 following CSO spills. Time series of *E. coli* concentrations at the five shellfish farming sites (Fig.
 14 1(c)) within the shellfish water following spills from five CSOs (red dots in Fig. 1(c)) are shown in
 15 Fig. 9 in log scale. Average bacterial concentrations at individual sites and the corresponding time of
 16 exposure at different concentrations are presented in Table 4. The *E. coli* maxima occurred within the
 17 first 24–36 h, which is the time when most discharges stopped spilling (Fig. 9(a)). Totnes STW SO
 18 continued to spill for a further 24 h, but this did not appear to affect bacterial levels in the shellfish
 19 water as much as the remaining four CSOs, which discharge near the shellfish farming sites. The
 20 highest *E. coli* concentrations were predicted at the farming sites 4 (1,600 cfu/100 ml) and 5 (2,500
 21 cfu/100 ml) (Figs 9(e) and 9(f), respectively). These sites are the closest to Galmpton PS SPS CSO,
 22 which is a very significant source of contamination locally in terms of discharge volume (Table 2).

1 Scenario 1 represents the highest average and instantaneous *E. coli* concentrations at the shellfish
2 farming sites (Fig. 9, Table 5) and corresponds to high river discharge volumes and neap tide. Under
3 these conditions, the transport of *E. coli* to the upper reaches of the estuary on the flood tide is
4 constrained by the outflowing river discharges, and thus bacterial contamination persists for some
5 time in the shellfish water. During neap tides and high river discharges, salinity is at its lowest level
6 (Table 5), which reduces the *E. coli* decay rate and this might contribute to maintain elevated *E. coli*
7 concentrations in the water.

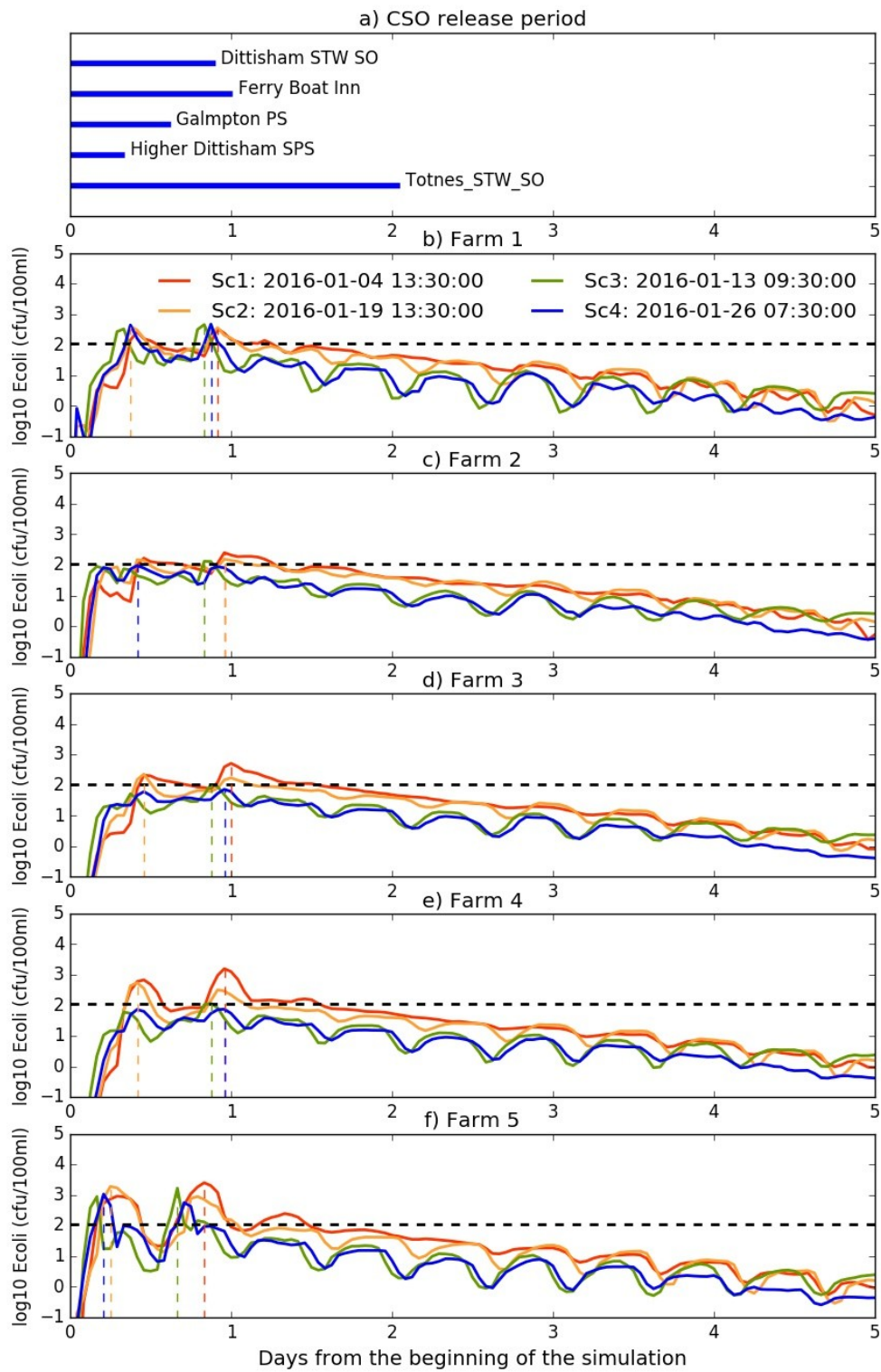
8 Scenario 2 shows higher peak and average concentrations than any other scenarios at all farming sites,
9 except Scenario 1. This is because under this scenario *E. coli* retention in the estuary is high, although
10 bacteria can still be transported further upstream along the estuary. In addition, the *E. coli* decay rates
11 in Scenario 2 were faster than in Scenario 1 due to higher salinities (Table 5).

12 Scenarios 3 and 4 represent high influence of marine waters in the estuary on spring tides and variable
13 river discharges. Scenario 3 represents lower river discharge and shows longer exposure time of *E.*
14 *coli* contamination in the SWPA than Scenario 4. The *E. coli* decay rates are also higher in Scenario 3
15 due to higher salinities. However, the average *E. coli* concentrations in both scenarios are relatively
16 similar (Table 5) with Scenario 3 showing slightly higher concentrations, probably due to longer
17 exposure time of bacterial contamination.

18 Consideration of these results in relation to the SWPAD and Regulation (EC) 627/2019 shows that an
19 *E. coli* concentration in the waters of 110 cfu/100 ml, which is equivalent to the Class B standard
20 under the regulation, was exceeded for several hours at all the farming sites under Scenarios 1 and 2
21 (the maximum is 22h under Scenario 1 at farming site 5). Under Scenarios 3 and especially 4, the
22 threshold is not achieved at some farming sites (Table 5). The *E. coli* threshold of 5 cfu/100 ml
23 (equivalent to the Class A standard) was exceeded for several days (up to 4 days at site 2 under
24 Scenario 1) under all scenarios at all the farming sites (Table 4).

25 The spatial distribution of *E. coli* concentrations in the estuary after 24 h (left column) and 48 h (right
26 column) of CSO spill start time under Scenarios 1–4 are shown in Fig. 10. The estuarine areas
27 affected by *E. coli* concentrations greater than 110 cfu/100 ml are enclosed by a white isoline in plots
28 (a) and (c). After 24 h of the beginning of the spills, the Class B standard was only exceeded under
29 Scenarios 1 and 2, especially at the shellfish farming areas (Figs. 10(a) and (10c)), with Scenario 2
30 showing a broader area of exceedance at this particular time. After 48 h of the beginning of the spills,
31 bacterial concentrations decreased below this threshold across the SWPA under all scenarios..

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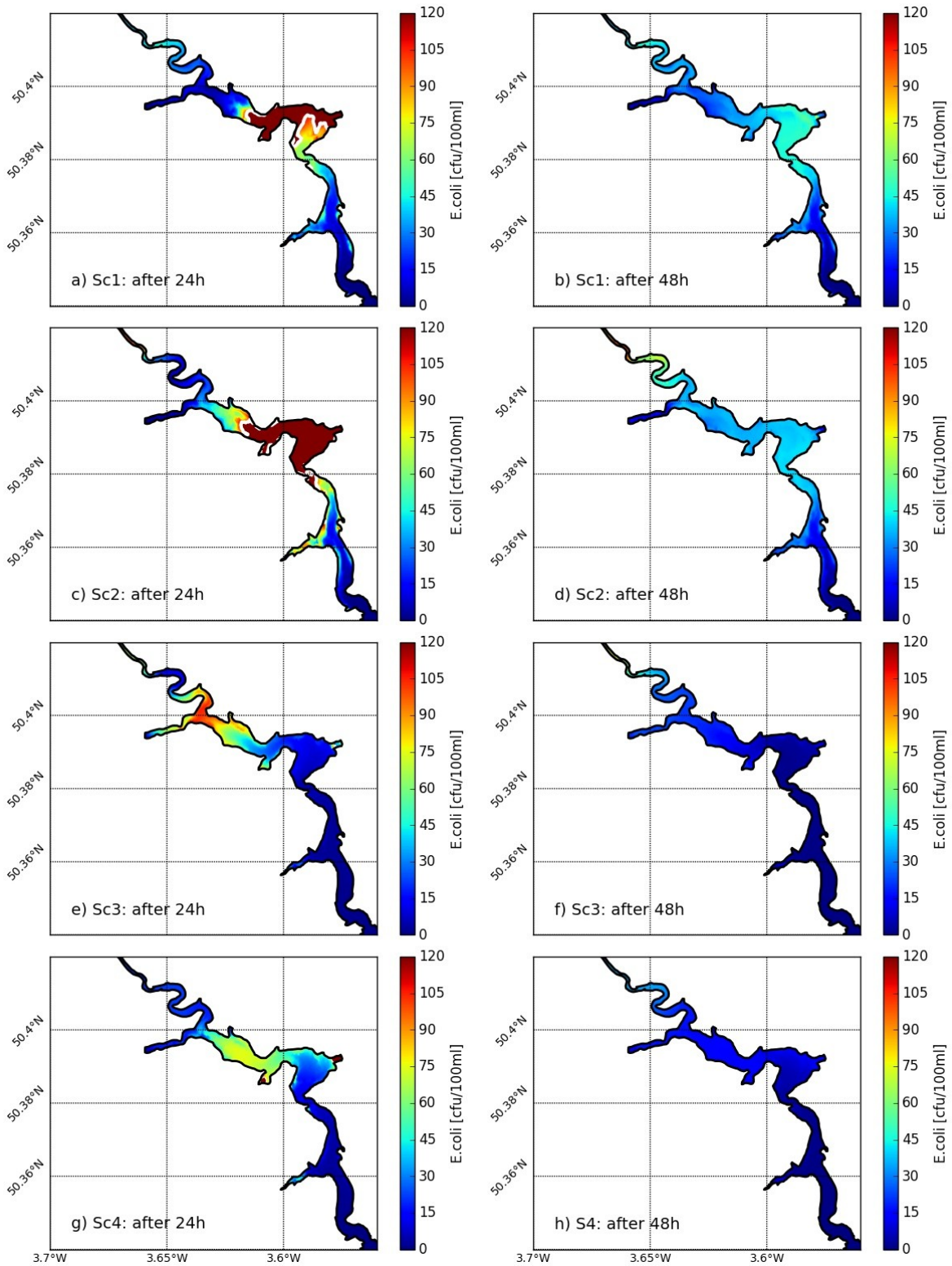
33 Fig. 9. Modelled *E. coli* concentrations at five shellfish farming sites (see Fig.1 (c)) for five days of
34 model simulations under Scenarios 1–4. The dashed black line represents the threshold of 110 cfu/100
35 ml in the water equivalent to the Class B standard in shellfish flesh. The vertical dashed lines

- 1 represent the time when peak *E. coli* concentration is achieved. Notice that at the farming site 2, the
- 2 peak concentration for Scenarios 1 and 2 is achieved at the same time.

1 **Table 5.** Mean salinities, *E. coli* concentrations and times of exposure to *E. coli* concentration greater than 110 cfu/100 ml and 5 cfu/100 ml at five shellfish
 2 farming sites during four contamination scenarios. *E. coli* in cfu/100 ml.

	Scenario 1				Scenario 2				Scenario 3				Scenario 4			
	Mean sal	Mean <i>E. coli</i>	Time (h) <i>E.</i> <i>coli</i> > 110	Time (h) <i>E.</i> <i>coli</i> > 5	Mean sal	Mean <i>E. coli</i>	Time (h) <i>E.</i> <i>coli</i> > 110	Time (h) <i>E.</i> <i>coli</i> > 5	Mean sal	Mean <i>E. coli</i>	Time (h) <i>E.</i> <i>coli</i> > 110	Time (h) <i>E.</i> <i>coli</i> > 5	Mean sal	Mean <i>E. coli</i>	Time (h) <i>E.</i> <i>coli</i> > 110	Time (h) <i>E.</i> <i>coli</i> > 5
Farm 1	9.21	39.58	12	86	18.63	37.66	8	87	21.63	23.78	5	68	15.43	21.33	4	59
Farm 2	6.45	41.68	14	95	16.77	34.64	5	93	20.53	17.31	2	74	12.81	16.29	0	63
Farm 3	8.13	54.43	21	90	18.00	32.44	7	90	20.97	13.40	0	73	14.34	12.045	0	61
Farm 4	9.56	93.74	22	90	19.09	43.28	10	88	21.98	13.43	1	7	15.34	11.99	0	58
Farm 5	12.39	144.32	22	91	21.79	101.07	13	87	24.01	39.12	7	61	18.05	35.29	5	57

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3 Fig. 10. Modelled concentrations of *E. coli* in the Dart Estuary. The left column shows
4 bacterial concentrations at 24 h after the beginning of the CSO spills; the right column
5 shows the same information at 48 h after the beginning of the spills. The white lines in
6 plots (a) and (c) represent the 110 cfu/100 ml *E. coli* concentration isoline.

1 3.2.2 *Effect of all storm overflows spilling concurrently (Scenario 5)*

2

3 A modelled scenario with all CSOs spilling concurrently showed *E. coli* concentrations increasing to a
4 peak level within approximately 24 h after spill start time and reducing thereafter under all scenarios
5 at most farming sites (Fig. 11). However, the *E. coli* concentration response to the spills differed
6 between sites. Farming sites 4 and 5 showed the highest *E. coli* concentrations. Under Scenario 5 at
7 site 2, *E. coli* concentrations peaked at 48 h after spill start. During the first three days after the spills,
8 *E. coli* concentrations exceeded for several hours a day the 110 cfu/100 ml threshold. Although most
9 CSOs stopped spilling within 24 h (except Totnes STW SO), *E. coli* concentration remained above
10 this limit for a few days at most farming sites.

11

12 3.2.3 *Effect of five storm overflows spilling at peak volume (Scenario 6)*

13

14 In this section, we analyse the modelling results of *E. coli* concentration in water at the five shellfish
15 farming sites with five CSOs spilling simultaneously at peak flow rates. Although this is a possible
16 scenario, it is likely that the duration of these peak events would be shorter than the one considered
17 here. Under peak flow discharge conditions, the water quality concentration equivalent to class B
18 standard (110 *E. coli*/100 ml) was frequently exceeded over four days after spills. *E. coli*
19 concentrations of 16,000 cfu/100 ml were predicted for the shellfish farming site 5 (Fig. 12(e)). At the
20 remaining four sites, the peak *E. coli* concentrations ranged from 200 cfu/100 ml to 4,500 cfu/100 ml.
21 These are around 10 times higher than the concentrations obtained under average discharge conditions
22 (low river discharge and neap tide, Scenario 2).

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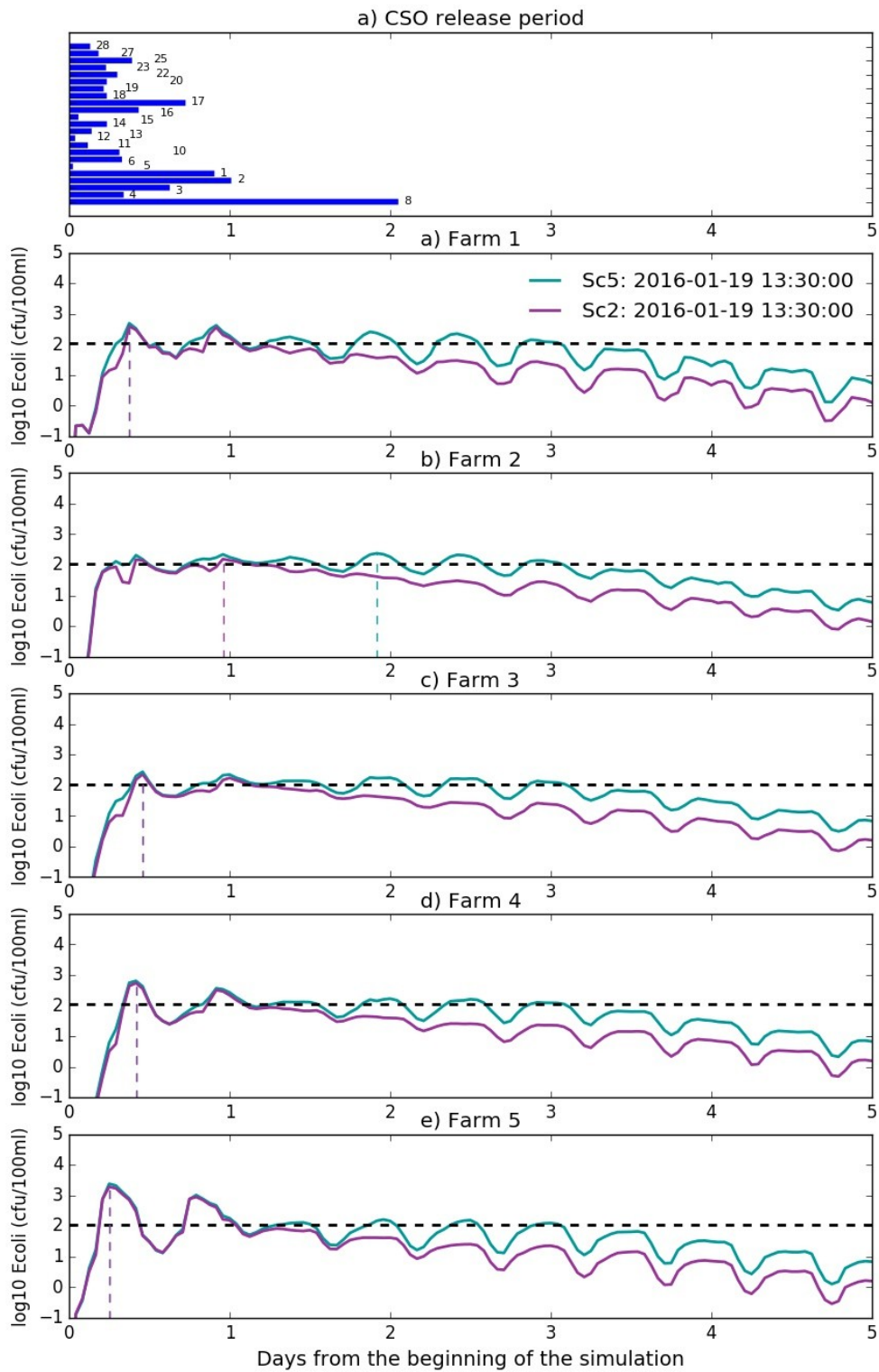
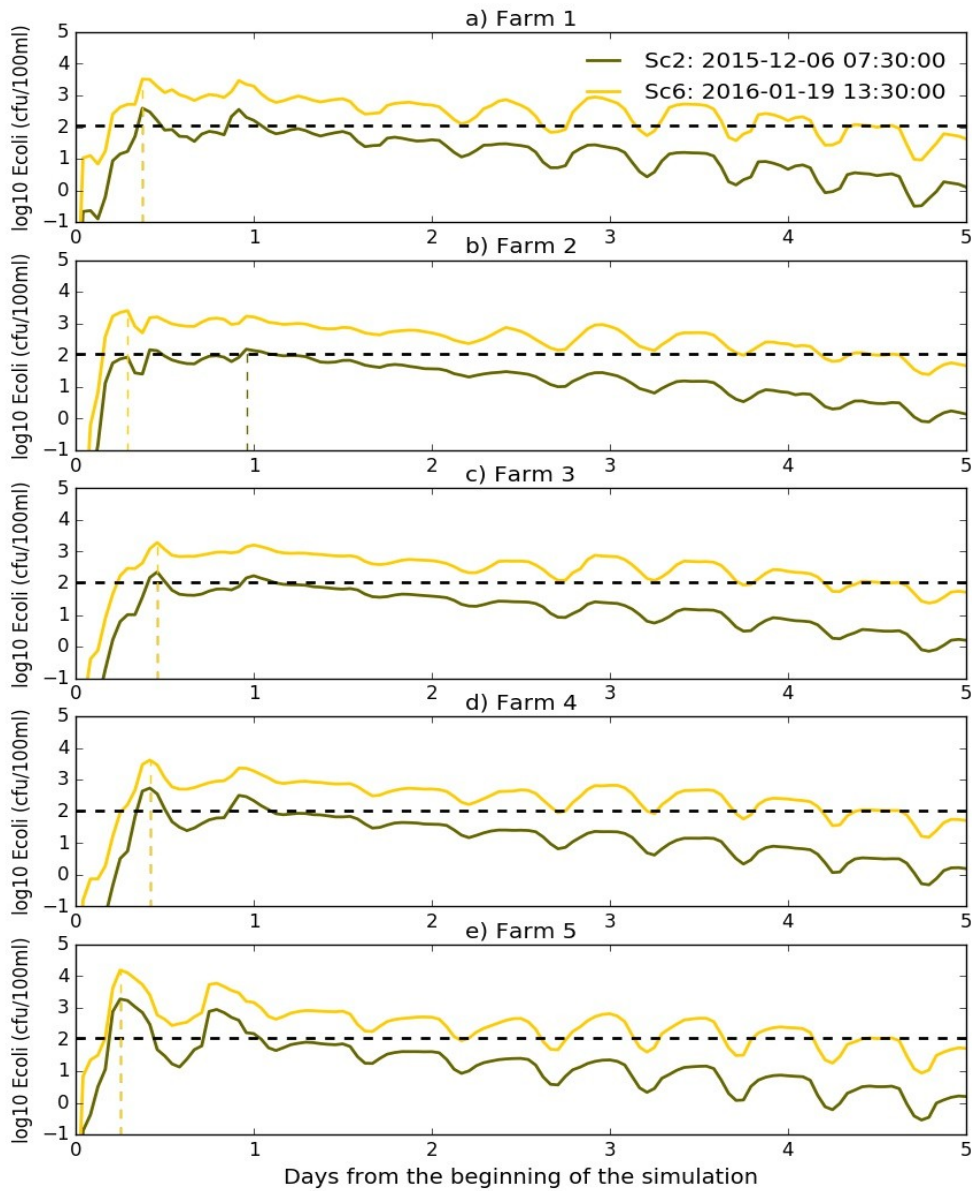


Fig. 11. Concentrations of *E. coli* at five shellfish farming sites considering all the CSOs spilling concurrently (Scenario 5) or only the five for which spill information is available (Scenario 2). Both scenarios are based on the same hydrodynamic conditions. The dashed black line represents the threshold of 110 cfu/100 ml in water (equivalent to the Class B standard in the shellfish). The vertical

1 dashed lines represent the time when peak *E. coli* concentration was achieved. Notice that the peak
2 was achieved at the same time for both scenarios at all farming sites, except for Farm 2.

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35 Fig. 12. Modelled concentrations of *E. coli* at five shellfish farming sites over five days following
36 sewage spills under peak flow discharge (Scenario 6) and average discharge (Scenario 2) conditions.
37 The same hydrodynamic conditions were considered in all model runs. The dashed black line
38 represents the threshold of 110 cfu / 100 ml in water (equivalent to the Class B standard in the
39 shellfish). The vertical dashed lines represent the time when peak *E. coli* concentration was achieved.
40 Notice that the peak was achieved at the same time for both scenarios at all farming sites, except at
41 Farm 2.

1 3.2.4 Effect of varying *E. coli* decay rates

2

3 This section reports on the effect of the two *E. coli* decay rates introduced in Section 2.5 on the model
4 results. Fig. 13 shows *E. coli* concentrations at the five shellfish farming sites predicted by the model
5 for the explicitly variable decay rate (EVDR; Scenario 2) and the implicitly variable decay rate
6 (IVDR; Scenario 7). The results show that the IVDR leads to much higher decay rates and hence to
7 lower concentrations of *E. coli* in water. At farming sites 2, 3 and 4, *E. coli* concentrations did not
8 exceed the threshold of 110 cfu/100 ml when the IVDR was considered. Bacterial concentrations for
9 IVDR were approximately four times lower than those for EVDR. At farming sites 1 and 5, the 110
10 cfu/100 ml threshold was exceeded when IVDR was used, but the *E. coli* concentrations were
11 approximately half of those obtained with an EVDR.

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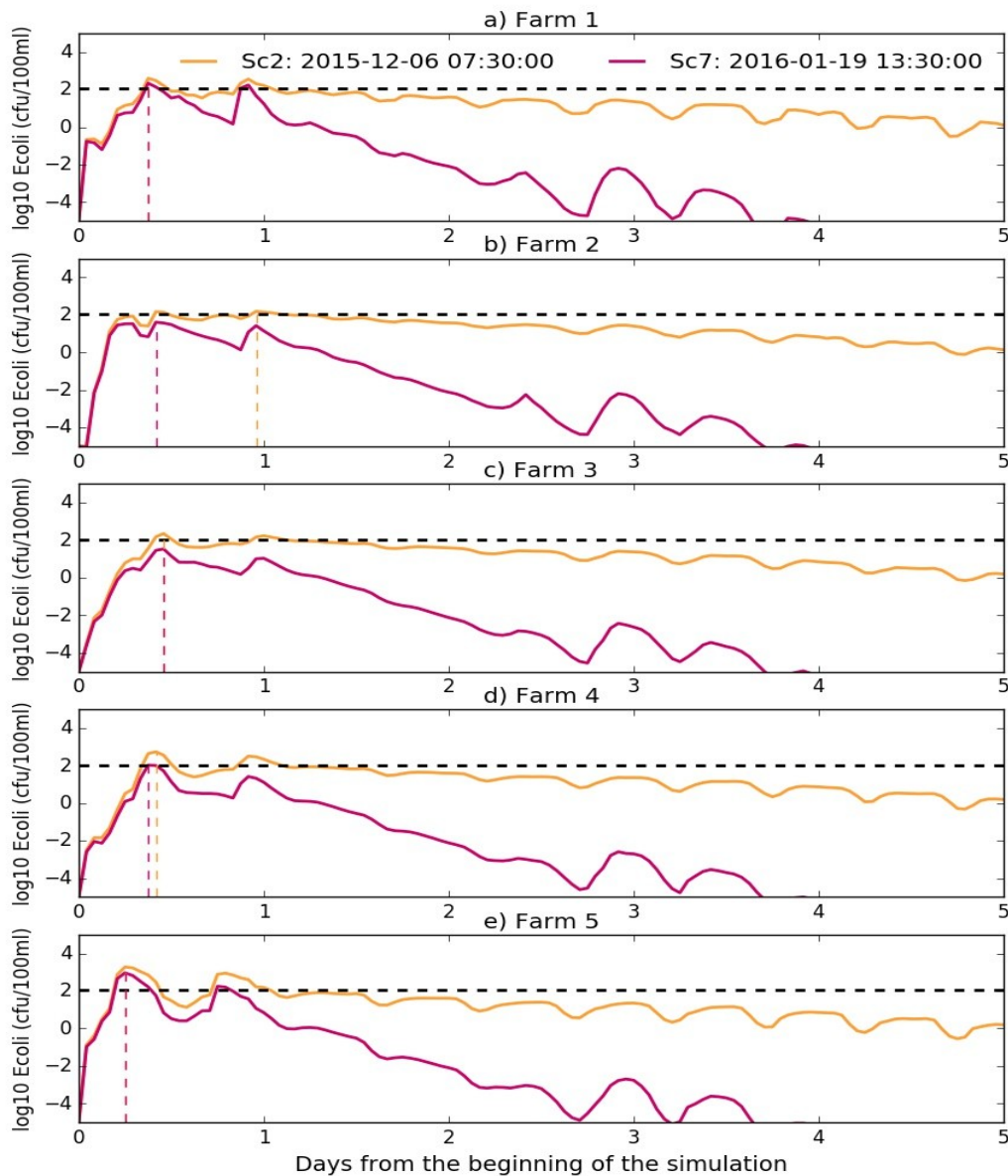
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27 13. Modelled concentrations of *E. coli* at five shellfish farming sites over five days of model

1 simulation under two bacterial decay rates. EVDR - explicitly variable decay rate (Scenario 2); IVDR
2 - implicitly variable decay rate (Scenario 7). The same hydrodynamic conditions were considered in
3 all model runs. The dashed black line represents the threshold of 110 cfu/100 ml in water (equivalent
4 to the Class B standard in the shellfish). The vertical dashed lines represent the time when peak *E.*
5 *coli* concentration was achieved. Notice that the peak was achieved at the same time for both
6 scenarios at all farming sites, except for Farm 2. The y axis has been truncated at a concentration of
7 10^{-5} cfu /100 ml (-5 in the log scale).

8

9 3.3. Comments on *E. coli* forecasting

10

11 Information on the association between rainfall and spill occurrence, frequency and duration can
12 inform decisions on management of microbiological contamination risks in SWPAs/shellfish farming
13 sites. In this study, we carried out an initial analysis of cumulative rainfall levels that may trigger
14 spills from the five CSOs considered in the model simulations. This is, those for which spill frequency
15 and duration data are available (Fig. 2). The rainfall data available for the period 2010–2016 described
16 in Section 2.8.4 were used for this analysis. A summary of results is compiled in Table 6.

17 The results indicate that, in general, a 3-day rainfall accumulation of 20–25 mm is likely to trigger a
18 spill. On the day the spill commences, an accumulation of between 12 and 20 mm is established and
19 rainfall remains greater than 20 mm for the duration of the spill at each CSO. Galmpton and Higher
20 Dittisham spill at higher cumulative rainfall (28.4mm and 26.2 mm, respectively), which may be
21 related to the greater capacity of these CSOs to store sewage (74 m³ and 20 m³, respectively). In
22 contrast, Dittisham and Ferry Boat Inn have lower storage capacity (7 m³ and 4.4 m³, respectively)
23 and therefore are more likely to spill at lower rainfall. The position at Totnes is in strong contrast with
24 the other four stations. Totnes serves an urban area with a population of ~8,000; the much higher
25 consented storage capacity of 3,520 m³ at Totnes STW reflects the much greater accumulations
26 possible over paved areas of the town.

27 To assess factors that may help “predict” the occurrence of a spill, we calculated the number of days
28 for which the accumulated rainfall was equal or exceeded the mean value during the spill (column 4 in
29 Table 6) and the percentage of time (days) that coincided with the occurrence of a spill. This analysis
30 shows that this was the case in more than 64% of cases, reaching 83% at Dittisham. The exception
31 was Higher Dittisham, where a value of less than 50% was obtained, although discontinuities exist in
32 the dataset of spill occurrence and duration in coincident periods where other CSOs register medium
33 to long term duration spills, (i.e. winter 2003 see Fig. 2 and Fig. 5). Applying the same analysis to
34 rainfall events > 50mm coincident with the occurrence of a spill results in a lowest value of more than
35 70% coincidence, and for some CSOs, 90% or even 100% coincidence (Dittisham).

36 The relationship between accumulated rainfall over the previous 3 days and the occurrence of a spill
37 was not very strong (<37% coincidence), except for Totnes STW SO, where 57% coincidence

1 between rainfall and spill occurrence was obtained. In summary, these results suggest that although
2 the occurrence of a spill is affected by rainfall falling on previous days (which takes up capacity of the
3 sewerage network to store sewage), the best indicator of spill occurrence seems to be rainfall of the
4 day.

1 **Table 6.** Summary statistics of association between cumulative rainfall and sewer overflow spills.

Name of sewer overflow	Accumulated rainfall (3 days prior to spill)		Accumulated rainfall (beginning of spill)		Accumulated rainfall (spill duration/day)		Percentage of rainfall events \geq Mean accumulated during spill	Percentage of rainfall events \geq Mean (resulting in a spill on the same day or within 2 days)	Percentage of rainfall events \geq 50 mm (resulting in a spill on the same day or within 2 days)
	Mean (mm)	95 th Conf. level	Mean (mm)	95 th Conf. level	Mean (mm)	95 th Conf. level			
Dittisham STW SO	20	2.4	14.5	1.6	26.2	6.4	83	37	100
Ferryboat Inn PS	23	3.3	12.7	1.7	22.1	5.6	64	35	70
Galmpton PS Higher	24	3.3	20.8	2.5	28.4	9.0	72	24	70
Dittisham Totness	26	3.9	18.4	2.4	20.6	2.7	48	17	70
STW SO	22.5	2.9	15.9	2.1	21.5	16.4	81	57	90

1 4. Discussion

2

3 A three-dimensional hydrodynamic model (TELEMAC 3D) was developed and successfully
4 applied to simulate the effect of the intermittent sewage discharges (CSOs) on the fate and transport of
5 *E. coli* in the Dart Estuary. Our interest lies in these bacteria because their presence indicates potential
6 human health risk from exposure to enteric pathogens via recreational contact with water and
7 consumption of contaminated shellfish. The model combines all the components that drive the
8 distribution of contamination and helps assess relative impact and associated health risk by simulating
9 different modelling scenarios. The wide range of riverine fluxes, variation in temperature and
10 salinities, and large tidal range in the estuary and the arrangements and operation of the CSOs
11 themselves in the catchment made this a complex problem to analyse.

12 Results show that the hydrodynamic conditions were key for the retention of *E. coli* in the
13 SWPA/farming sites, with neap tides and high river discharges leading to the highest *E. coli*
14 concentrations. Also, the distance from the CSOs to the farming sites, the duration and intensity of the
15 spills and whether the spills were concurrent were all shown to affect the resultant concentrations. *E.*
16 *coli* experiences a decay through time due to increased mortality, and this depends on many factors
17 (Campos et al. 2013). This non-conservative behaviour was implemented in the model through two
18 methods, which resulted in very different *E. coli* concentrations at the farming sites: the more
19 sophisticated EVDR method results in higher concentrations and relies on the application of a fully
20 baroclinic 3D model to acquire the fully resolved temperatures, salinities and irradiation; and the less
21 complex IVDR, that leads to lower concentrations, and parametrises decay rates based on varying the
22 rate in periods of day and night, which may be suited to a simpler, 2D modelling approach. A model
23 of this type was applied by Robins et al. (2019) on the river dominated part of the Dee Estuary (North
24 Wales) to examine the transport of adenovirus through the river system. This research did not include
25 ephemeral storm discharge from CSOs along the river but used sampling and stochastic methods to
26 gauge the extent of dispersal which were highly dependent upon river and tidal conditions. Examples
27 of scenario-based modelling approaches similar to ours involving CSOs in other rivers can be found
28 in Even et al. (2007) and Björklund et al. (2018).

29 Unfortunately, a sustained dataset of *E. coli* concentrations in water does not exist for the Dart
30 Estuary, which hindered the task of the model calibration and validation and constrained the
31 operational use of the model. The only observational data available at Waddeton. were approximately
32 measured between farming sites 3 and 4 in Figure 1 (c) and are not likely to reflect current conditions
33 in terms of microbiological pollution in the estuary. Comparable concentrations to those detected in
34 the waters were only obtained when the CSOs were spilling at peak flow conditions.

35 The model presented here focuses on scenarios designed based on available data and/or previous
36 modelling results, and potentially constitutes a valuable tool for managers to understand the
37 conditions under which water/shellfish standards may be exceeded and for how long those

1 exceedances are likely to occur, as well as to plan actions to improve the water quality of the Dart
2 Estuary. However, some improvements to make the model more realistic and adapted to the local
3 conditions would be needed before the model scenario outputs presented here can be used with
4 confidence by water resource managers in the Dart. Some of these are: a) including wind forcing,
5 since it influences the hydrodynamic circulation; b) explicitly calculating the 3D temperatures (now
6 considered constant in time and space), which would result in a better characterization of the water
7 column structure and would allow to account for the variability in temperature in the *E. coli* decay
8 rates and c) acquiring further monitoring data to validate the model, parametrise *E. coli* decay and
9 characterise CSO discharge effects over time. Given all these uncertainties, the results shown here
10 should be taken with caution, especially with regard to the exceedance of regulatory levels.
11 Nevertheless, it seems that for average CSO spill frequency and duration, it is likely that the 110
12 cfu/100 ml concentration in water corresponding to the Class B standard of Regulation (EC) No
13 627/2019 will be exceeded for several hours in the Dart Estuary. This generally occurs within the first
14 day after the beginning of the spill and especially at the shellfish farming sites closer to the CSOs (i.e.
15 site 5) and under certain hydrodynamic conditions, even when faster decay rates are considered
16 (IVDR). For the slower decay rates (EVDR) the period of exceedance would be longer. The *E. coli*
17 concentration of 5 cfu/100 ml (equivalent to Class A standard of the regulation) is likely to be
18 exceeded during the first day after CSO spill start times (when IVDR decay rates are considered),
19 increasing up to 4 days (depending on the hydrodynamic conditions and location of farming sites in
20 relation to the CSOs) when EVDR decay rates are considered. These results are consistent with the
21 lag time of 3–4 days between the occurrence of rainfall and *E. coli* maxima detected at the shellfish
22 farming sites in the Dart, as reported by Campos et al. (2011).

23 A preliminary study was also undertaken to characterize the rainfall conditions that would lead to
24 the occurrence of CSO spills in the Dart Estuary. Model outputs demonstrated that spills generally
25 occurred with average rainfall > 20 mm, that variability in response to rainfall event exists between
26 individual CSOs, and that the response of these discharges to a rainfall event is faster during the first
27 day of the event than when accumulated rainfall over previous days is considered. The absence of a
28 time series of spill data did not allow for a full characterization of CSO spill variability in relation to
29 the magnitude of rainfall. Moreover, the rainfall data were obtained from ECMWF ERA-Interim,
30 which has a relatively coarse resolution of 0.75 degrees and therefore cannot provide summary
31 statistics for specific parts of the catchment. Undoubtedly, rain gauge data in the vicinity of the CSOs
32 would have provided a better characterization. In addition to the rainfall, other factors such as the
33 catchment area, population connected to the network or storage capacity of each overflow influence
34 spill frequency and magnitude. Therefore, an operational forecasting system for water quality in the
35 Dart Estuary would ideally couple a Storm Water Management Model (SWMM) to represent the
36 catchment and sewer system, with a hydrodynamic model such as the one reported here. Some

1 examples of these coupled models can be found in the literature (De Marchis et al., 2013; Locatelli et
2 al., 2019).

3 4 **5. Conclusions**

5
6 The three-dimensional hydrodynamic model of the Dart Estuary reported in this paper can
7 reproduce water levels and current velocities, as well as periods of stratification and mixing generated
8 by tides and varying river discharge in the estuary. The model was used to investigate the fate and
9 transport of *E. coli* from CSOs. The model incorporates constant flow volumes and average spill
10 duration to simulate a set of discharge scenarios according to varying hydrodynamics, spill frequency
11 and duration and *E. coli* decay rates. From these modelling scenarios, it was concluded that:

- 12
13 • High volumes of river discharge at neap tides lead to larger areas of *E. coli* contamination
14 impact in the SWPA and longer exposure of shellfish to peak *E. coli* concentrations due to a
15 combination of high retention and, to a lesser extent, lower *E. coli* decay rates.
- 16 • In extreme rainfall events, if all CSOs spill at average flow volumes and are coincident in
17 time, the peak *E. coli* concentrations in the water column do not increase significantly in
18 magnitude when compared to a scenario of spills from a reduced number of CSOs, i.e. those
19 located near the shellfish farming sites. However, elevated levels of *E. coli* in the water,
20 exceeding the equivalent of class B standard in shellfish flesh, are likely to occur over longer
21 periods of time at the farming sites.
- 22 • During peak spill volumes, *E. coli* concentrations in the SWPA can increase up to 10 times the
23 concentration predicted under average flow conditions.
- 24 • Rainfall events greater than 20 mm are likely to trigger spills from CSOs which will affect
25 microbiological contamination in the SWPA.
- 26 • Model outputs for time periods when tidal and river flow conditions dictate elevated *E. coli*
27 concentrations combined with rainfall forecasts for the Dart catchment can provide a good
28 indication of microbiological contamination risk in shellfish farming sites and recreational
29 waters.

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33
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4 **Author contributions**

5 L.G., C.C., S.K., A.Y. and J.B. designed the research and drafted the manuscript; L.G. and J.B.
6 analysed the data and conducted the modelling studies.

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