


Modelling the quality of bathing waters in the Adriatic Sea

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Abstract: The aim of this study is to develop a relocatable modelling system able to describe the microbial contamination that affects the quality of coastal bathing waters. Pollution events are mainly triggered by urban sewer outflows during massive rainy events, with relevant negative consequences on the marine environment and tourism and related activities of coastal towns. A finite element hydrodynamic model was applied to five study areas in the Adriatic Sea, which differ for urban, oceanographic and morphological conditions. With the help of transport-diffusion and microbial decay modules, the distribution of *Escherichia coli* was investigated during significant events. The numerical investigation was supported by detailed in situ observational datasets. The model results were evaluated against water level, sea temperature, salinity and *E. coli* concentrations acquired in situ, demonstrating the capacity of the modelling suite in simulating the circulation in the coastal areas of the Adriatic Sea, as well as several main transport and diffusion dynamics, such as riverine and polluted waters dispersion. Moreover, the results of the simulations were used to perform a comparative analysis among the different study sites, demonstrating that dilution and mixing, mostly induced by the tidal action, had a stronger effect on bacteria reduction with respect to microbial decay. Stratification and estuarine dynamics also play an important role in governing microbial concentration. The modelling suite can be used as a beach management tool for improving protection of public health, as required by the EU Bathing Water Directive.

Keywords: numerical model; bathing water; faecal pollution; Adriatic Sea

1. Introduction

Microbiological contamination of marine water bodies is one of the biggest environmental concerns in coastal zones subjected to rapid population growth [1]. Faecal bacteria (e.g. *Escherichia coli* and Enterococci) originating from human faeces and organic waste in the sewage, as well as animal faeces in run-off, disposed of in the water bodies without any sanitation systems, constitute essential sources for the marine environmental contamination [2,3]. Consequently, human health can be seriously endangered and a bad bathing water quality can have adverse effects on the tourist industry and many recreational and economic activities [4]. Storm runoff has become one of the major

29 sources of pollutant loading, including pathogens, pesticides, heavy metals and nutri-
30 ents to the coastal recreational waters [5,6]). In the last decades, extraordinary strong
31 storm events have become more and more regular in many areas with the prediction of
32 further rise in their frequency [3]. During these uncontrolled storm events, combined
33 sewer overflows (CSOs) discharge high concentrations and loads of *E. coli* and intestinal
34 enterococci bacteria in the receiving water bodies, where faecal bacteria concentrations
35 can easily exceed the bathing water quality standards [7].

36 At the European level the Bathing Water Directive [BWD 8] and the corresponding
37 transposition law within each EU nation, define threshold levels for intestinal enterococci
38 and *Escherichia coli* concentrations to prevent the health risks associated with bathing
39 in marine and freshwaters. The BWD establishes the guidelines for bathing water
40 monitoring and classification, the management, and the provision of information to the
41 public. In an efficient water quality preservation program, mitigating the microbiological
42 contamination of marine waters requires an integrated assessment, whenever necessary,
43 to enable decision-makers to adopt adequate mitigation actions and to explore the
44 consequences of various management options for protecting public health [1,9].

45 Limiting the exposure to polluted water requires a quantitative understanding of
46 storm runoff impact on coastal water quality and predictive models that can forecast
47 the water quality for timely management decisions [10]. In this regard, oceanographic
48 models constitute a useful tool for determining the concentration of faecal bacteria in
49 nearshore bathing waters [11–13] and to calculate the risk for human health caused by
50 microbial pollution [14,15]. Simulation of water circulation and the transport processes
51 affecting coastal areas requires the use of numerical models at high spatial resolution
52 capable of representing complex morphological and bathymetric features as well as
53 several anthropogenic constructions (piers, artificial reefs, breakwaters, jetties) present
54 along the coast. Moreover, the correct representation of the coastal dynamics requires
55 high temporal resolution to describe processes occurring at a short time scale, like tidal
56 fluctuation and flash river floods [e.g. 16]. The simulation of nearshore water quality
57 during and after sewer overflows requires high-resolution models as well as detailed
58 information concerning water discharge and microbial concentration input values [7].
59 The last requirement is crucial in many coastal sites, where the sewer outflows are
60 not continuously monitored or where illegal sewer connectors discharge into the sea.
61 Therefore, all coastal simulations must be supported by an adequate dataset for their
62 implementation and validation.

63 By understanding the dynamics associated with faecal contamination, it is possible
64 for managers and policymakers to incorporate those findings to develop sound sampling
65 strategies and attenuation measures in order to avoid bathing area closures for prolonged
66 and unnecessary periods of time. Normally, the application of numerical dispersion
67 models is used to support the traditional monitoring methods based on field observations
68 and laboratory analysis in order to link information concerning the hydrodynamic
69 circulation, environmental parameters and the microbiological features of an area [7,17].

70 Most of the previous modelling studies have been limited to one coastal system
71 and they lack an integrated, comprehensive evaluation in different environments. In this
72 study, we describe a relocatable modelling system for assessing microbial pollution in
73 coastal areas which consists of a hydrodynamic model, a transport and diffusion module
74 and a microbial decay module. The adopted approach realises a seamless transition
75 between different spatial scales, from the river mouth to the open sea, and adopts a high
76 spatial and temporal resolution of the forcing and boundary conditions that drive the
77 simulations. The model is evaluated against observations in the coastal areas, illustrating
78 the capability of this tool in simulating the water circulation as well as the dispersion
79 and decay of microbial pollutants. The model has been applied to five different coastal
80 areas located on both the western and the eastern sides of the Adriatic Sea, a region
81 of the Mediterranean sea considered a very sensitive area due to the heavy organic,
82 eutrophication substances and other pollutants discharged through the main rivers

83 [17–20]. Alongside these forms of pollution, the problem of microbial contamination is
84 also particularly relevant for the Adriatic Sea coast, where several urban settlements,
85 popular bathing locations and tourist centres are located [20–24].

86 1.1. Study areas

87 This study focuses on five target coastal areas located in the Adriatic Sea, an 800-km-
88 long, 150-km-wide elongated semi-enclosed basin interacting with the Mediterranean
89 Sea through the Otranto Strait in the southern part (Fig. 1). The main forcings of the
90 Adriatic basin circulation are the wind (influenced by the complex local orography
91 and small scale processes), the strong buoyancy resulting from the freshwater inputs
92 injected by the rivers and the tidal waves propagating from the Mediterranean Sea. The
93 general surface circulation of the Adriatic Sea may be described as a large-scale cyclonic
94 meander, with a northerly flow along the eastern coast and a southerly return flow along
95 the western coast and a double gyre configuration in the central and southern parts of
96 the basin [25,26].

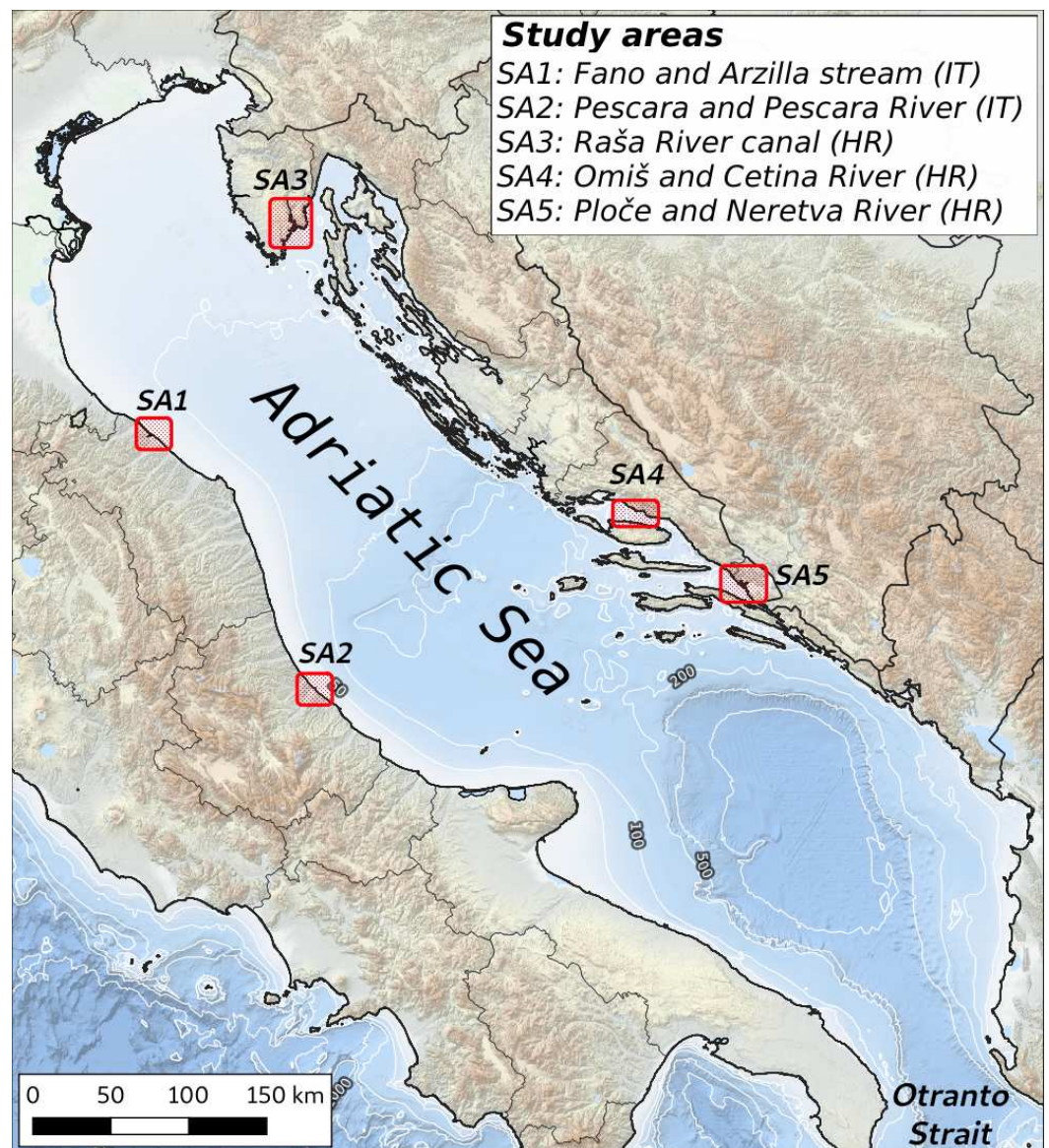


Figure 1. The Adriatic Sea with the red rectangles indicating the five coastal study areas. Background: EMODNet bathymetry [27].

97 As in other countries, increasing population and rapid urban development along
 98 the coastline of the Adriatic sea have caused a dramatic increase in sewage discharge
 99 into rivers and the sea [20,22]. Most of this sewage has undergone no more than primary
 100 treatment and threatens the health of aquatic ecosystems and directly and indirectly
 101 affects human health and recreational opportunities along with the coast [18,28,29].
 102 Bathing waters and recreational activities are a resource of great economic and environ-
 103 mental importance and their safety is a primary goal in the management of the coastal
 104 area. Even if Italy has adopted criteria highly restrictive in terms of quality of bathing
 105 waters and monitoring, recent studies underlined the persistence of many many critical
 106 situations in various parts of the Adriatic coasts [17,19,23] and consequently, bathing
 107 has been prohibited at different points.

108 A wide variety of coastal environments are present along the coastline of the Adriatic
 109 Sea. The eastern and western sides of the Adriatic Sea greatly differ in appearance:
 110 the western coast is largely sedimentary, with mild sloping and sandy beaches, while
 111 the eastern coast is composed of many islands and headlands and is generally rugged
 112 and rocky. All areas investigated in this study are coastal zones located near urban
 113 settlements and influenced by the discharge of a river collecting wastewaters from the
 114 local sewerage system. As indicated in Fig. 1, they are:

- 115 • SA1: the coast of Fano at the mouth of the Arzilla stream (Marche region, Italy);
- 116 • SA2: the coast of Pescara at the mouth of the Pescara River (Abruzzo region, Italy);
- 117 • SA3: the fjord-like system of the Raša River (Istria region, Croatia);
- 118 • SA4: the coast of Omiš at the Cetina River mouth (Split-Dalmatia region, Croatia);
- 119 • SA5: the Ploče coast with the Neretva Estuary (Dubrovnik-Neretva region, Croatia).

120 These environments give a representative picture of the different coastal systems
 121 situated around the Adriatic basin covering a wide range of urban, hydrological and
 122 oceanographic conditions. An overview of the study sites is provided in Fig. 2 and a
 123 comprehensive description of their characteristics is reported in section 2.2.

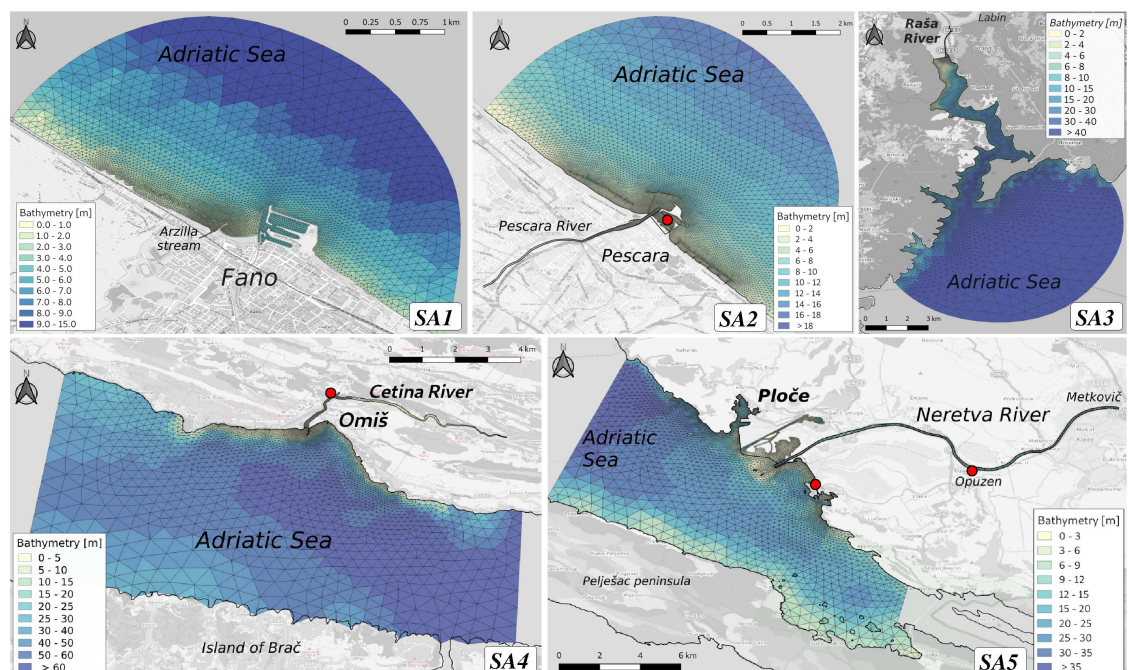


Figure 2. Numerical grids with the bathymetry superimposed of the five study areas. The red dots mark the location of the water level monitoring stations at the Pescara, Omiš-Cetina and Ploče-Neretva study areas. Background: OpenStreetMap.

124 2. Materials and Methods

125 2.1. Model description

126 The modelling framework presented here is based on the System of Hydrodynamic
127 Finite Element Modules [SHYFEM, 30] code, an open-source unstructured ocean model
128 for simulating hydrodynamics and transport processes at very high resolution. The
129 modelling suite consists of:

- 130 • a 3D hydrodynamic model, that describes currents and mixing of water mass in the
131 system;
- 132 • a transport and dispersion module, that simulates the dispersion of solute and
133 microorganisms through the system;
- 134 • a microbial decay module, which defines the decay of microorganisms considering
135 various environmental conditions.

136 The horizontal discretization of the state variables is carried out with the finite
137 element method, with the subdivision of the numerical domain in triangles varying in
138 form and size. Such a method has the advantage of representing in detail complicated
139 bathymetry and irregular boundaries in coastal areas. Thus, it can solve the combined
140 large-scale oceanic and small-scale coastal dynamics in the same discrete domain by
141 using unstructured meshes. In the following sections, the single modules are described
142 in detail.

143 2.1.1. The hydrodynamic model

144 The 3D hydrodynamic finite element model is based on the solution of primitive
145 equations and previously applied on several transitional environments, coastal and
146 shallow basins. The model has been already applied to simulate hydrodynamics in the
147 Mediterranean Sea [31], in the Adriatic Sea [32] and in several coastal systems [33, and
148 references therein]. [34] demonstrated the good performance of the SHYFEM model in
149 simulating water levels, currents, salinity, and water temperature in the Adriatic Sea.

150 The hydrodynamic model is the “engine” that transports and mixes all ecosystem
151 constituents, including the water itself. The model solves in a 3D formulation the shallow
152 water equations, which for an arbitrary vertical layer l are the following:

$$153 \frac{\partial U_l}{\partial t} + u_l \frac{\partial U_l}{\partial x} + v_l \frac{\partial U_l}{\partial y} - fV_l = -gh_l \frac{\partial \zeta}{\partial x} - \frac{gh_l}{\rho_0} \frac{\partial}{\partial x} \int_{-H_l}^{\zeta} \rho' dz - \frac{h_l}{\rho_0} \frac{\partial p_a}{\partial x} \quad (1a)$$

$$+ \frac{1}{\rho_0} \left(\tau_x^{top(l)} - \tau_x^{bottom(l)} \right) + \frac{\partial}{\partial x} \left(A_H \frac{\partial U_l}{\partial x} \right) + \frac{\partial}{\partial y} \left(A_H \frac{\partial U_l}{\partial y} \right) + \frac{\partial}{\partial z} \left(A_v \frac{\partial U_l}{\partial z} \right)$$

$$154 \frac{\partial V_l}{\partial t} + u_l \frac{\partial V_l}{\partial x} + v_l \frac{\partial V_l}{\partial y} + fU_l = -gh_l \frac{\partial \zeta}{\partial y} - \frac{gh_l}{\rho_0} \frac{\partial}{\partial y} \int_{-H_l}^{\zeta} \rho' dz - \frac{h_l}{\rho_0} \frac{\partial p_a}{\partial y} \quad (1b)$$

$$+ \frac{1}{\rho_0} \left(\tau_y^{top(l)} - \tau_y^{bottom(l)} \right) + \frac{\partial}{\partial x} \left(A_H \frac{\partial V_l}{\partial x} \right) + \frac{\partial}{\partial y} \left(A_H \frac{\partial V_l}{\partial y} \right) + \frac{\partial}{\partial z} \left(A_v \frac{\partial V_l}{\partial z} \right)$$

$$154 \frac{\partial \zeta}{\partial t} + \sum_l \frac{\partial U_l}{\partial x} + \sum_l \frac{\partial V_l}{\partial y} = 0 \quad (1c)$$

155 with U_l, V_l the horizontal transport at each layer (integrated velocities), f the Coriolis
156 parameter, p_a the atmospheric pressure, g the gravitational acceleration, ζ the sea level,
157 ρ_0 the average density of sea water, $\rho = \rho_0 + \rho'$ the water density, τ the internal stress
158 term at the top and bottom of each layer, h_l the layer thickness, H_l the depth at the
159 bottom of layer l . The Smagorinsky’s formulation [35,36] is used to parameterize the
160 horizontal eddy viscosity (A_H). For the computation of the vertical viscosities (A_v) a
161 turbulence closure scheme was used. This scheme is an adaptation of the k- ϵ module of
162 the General Ocean Turbulence Model described in [37].

163 Velocities are computed in the centre of the grid element, whereas water levels
 164 are computed at element vertices (nodes). The model uses a semi-implicit algorithm
 165 for integration over time, which has the advantage of being unconditionally stable for
 166 gravity waves, bottom friction and Coriolis terms, and allows transport variables to
 167 be solved explicitly. The model adopts automatic sub-stepping over time to enforce
 168 numerical stability for advection and diffusion. A more detailed description of the model
 169 equations and the discretization method is given in [30,38].

170 2.1.2. The transport and diffusion module

171 The 3D Eulerian transport-diffusion model solves the following equation:

$$\frac{\partial C_l}{\partial t} + u_l \frac{\partial C_l}{\partial x} + v_l \frac{\partial C_l}{\partial y} + w_l \frac{\partial C_l}{\partial z} = \frac{\partial}{\partial x} \left(K_H \frac{\partial C_l}{\partial x} \right) + \frac{\partial}{\partial y} \left(K_H \frac{\partial C_l}{\partial y} \right) + \frac{\partial}{\partial z} \left(K_V \frac{\partial C_l}{\partial z} \right) + E - K_d C_l \quad (2)$$

172 where C_l is the concentration at the layer l of the solute (tracer, salinity, water tempera-
 173 ture); u_l , v_l and w_l are the horizontal and vertical component of the currents; K_H and K_V
 174 are respectively the horizontal and vertical turbulent diffusion coefficients calculated
 175 the former with the Smagorinsky formulation [35] and the latter calculated by the $k-\epsilon$
 176 turbulence closure model; E is the sink/source term; K_d is the decay rate. The transport
 177 and diffusion equation is solved with a first-order explicit scheme based on the total
 178 variational diminishing scheme.

179 In the case of salinity, the source/loss term E represents the difference between evap-
 180 oration and precipitation through the water surface. The evaporation rate is determined
 181 by the bulk aerodynamic transfer method [39] using measurements of air temperature,
 182 relative humidity, wind speed, air pressure and simulated water temperature. In case
 183 of water temperature, E represents the heat source through the water surface $Q/\rho c_w h_l$,
 184 where ρ is the water density, c_w is the specific heat of water ($c_w=3991 \text{ J kg}^{-1} \text{ }^\circ\text{C}^{-1}$) and
 185 h_l is the depth of fluid layer. Q is the heat flux between the atmosphere and the sea
 186 computed by the energy-radiation balance considering short and longwave radiation,
 187 heat flux generated by the evaporation-condensation process and heat flux generated by
 188 convection and conduction process.

189 2.1.3. The microbial decay module

In marine coastal environments, the fate of free-living faecal bacteria in the water
 column is approximated as an Eulerian tracer (term C in equation 2) subjected to dilution
 and a decay relationship. Following Ostoich *et al.* [40], in this study, the decay rate (K_d)
 is considered variable in space and time as a function of the environmental conditions,
 e.g. water temperature, salinity, water turbidity and UV radiation. To implement this
 equation, we followed that proposed by Chapra [41], where the total loss rate for bacteria
 can be read as:

$$K_d = K_{base} + K_{solar} \quad (3)$$

where K_{base} is the base mortality rate and K_{solar} is the loss rate due to solar radiation.
 Other significant processes, like settling and adsorption in suspended particulate matter,
 are not considered in this study. The K_{base} term [42] for sea waters can be presented as:

$$K_{base} = (0.8 + 0.02S)1.07^{(T-20)} \quad (4)$$

190 where S_l is the salinity and T_l is the water temperature.

The term K_{solar} is proportional to the surface light energy (I_0) through a constant α , approximately equal to 1 [43]. Using the Lambert-Beer law to model the exponential decay of light in a well-mixed layer, K_{solar} can be calculated as:

$$K_{solar} = (I_0/k_e H)(1 - \exp^{-k_e H}) \quad (5)$$

191 where k_e is the light extinction coefficient in the water and H is the water depth. Accord-
 192 ing to Feitosa *et al.* [44], the approach of Mancini [42] is recommended for estimating
 193 bacterial decay rates under day and nighttime conditions and considering the combined
 194 influences of temperature and salinity. The simulated bacterial concentration is highly
 195 sensitive to the prescribed decay rate [13, and reference therein], but in highly dynamic
 196 environments, dilution has generally a higher effect on bacterial reduction than decay
 197 rate Eregno *et al.* [14], Madani *et al.* [45]. The bacteria decay module was tested using
 198 ambient values of solar radiation, water temperature and salinity registered in Fano
 199 in summer 2019. The obtained decay rate (T90, defined as the time at which 90% of
 200 the bacterial population is no longer detectable) is presented in Fig. 3. The decay rate
 201 varies from 3 h during the night to a peak daily value of more than 30 h with clear sky
 202 conditions. The decay rate increases with the salinity and with the water temperature.
 203 Such T90 values were within the proposed ranges of Feitosa *et al.* [44] and Ostoich *et al.*
 204 [40].

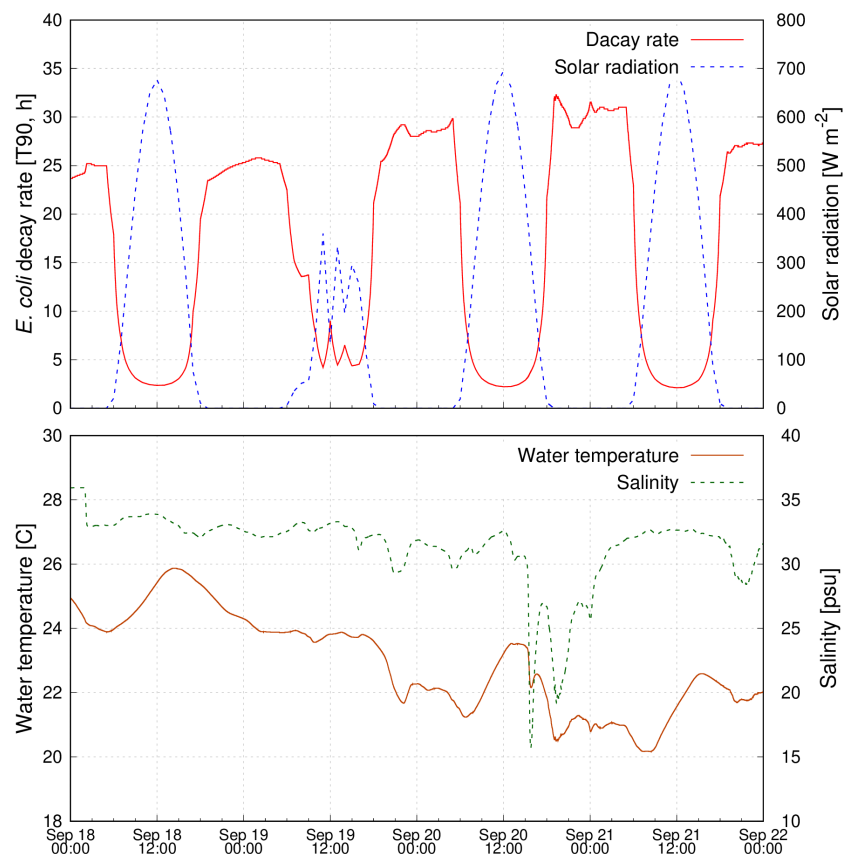


Figure 3. Variation of the *E. coli* decay rate (T90) as a function of incident solar radiation (top panel), water temperature and salinity (bottom panel) in the period 18-22 September 2019 in Fano.

205 The decay equations have been integrated into the model for each node and in the
 206 middle of each layer, considering a value of 1 for the extinction coefficient increasing
 207 progressively the total depth of the column from surface to bottom. At each time step,
 208 the model simulates the dispersion of faecal bacteria from the sewer outflow into the
 209 coastal waters.

210 2.2. Model implementation

211 The model runs in a full 3-D baroclinic mode with the water column discretized in
212 zeta or hybrid (mixed sigma and zeta) layers with varying thickness (see the site-specific
213 settings illustrated below). We performed several numerical experiments for simulating
214 hydrodynamic conditions and dispersion of *E. coli* in the five study sites. The discharge
215 of the sewerage outlet is simulated with an Eulerian approach, where the concentration
216 of *E. coli* is prescribed in outflow and the impact of the concentration is evaluated on the
217 coast. The simulations were forced:

- 218 • at the sea open boundary by sea temperature, salinity, water level and currents
219 conditions obtained from the TIRESIAS operational system of the Adriatic Sea
220 [34]. Such an unstructured oceanographic model reproduces in detail the general
221 circulation in the Adriatic Sea, as well as several relevant coastal dynamics, like
222 tidal amplification, saltwater intrusion, storm surge and riverine water dispersion;
- 223 • at the sea surface by meteorological data (air temperature, solar radiation, humidity,
224 cloud cover, mean sea level pressure, wind speed and direction) from the high-
225 resolution MOLOCH model [46]. The MOLOCH model is implemented with a
226 horizontal grid spacing of 1.25 km, and with 60 atmospheric levels and 7 soil levels
227 and provides the meteorological parameters at hourly frequency;
- 228 • at the river boundary by water discharge timeseries computed from observed water
229 levels through calibrated stage-discharge relationships;
- 230 • at the pollutant sources by bacteria concentration and water volume according to
231 the available site-specific data.

232 For the free surface, a water flux is used containing evaporation minus precipitation
233 and river discharge. For computing the water temperature, the air-sea heat fluxes are
234 parameterised by the Coupled Ocean-Atmosphere Response Experiment (COARE) 3.0
235 bulk algorithm [47]. Also, the drag coefficient for the momentum transfer of wind in
236 the hydrodynamic model is computed according to the COARE 3.0 bulk formulae [47].
237 The bottom drag coefficient is computed using a logarithmic formulation via bottom
238 roughness length, set homogeneous over the whole model domain to a value of 0.01
239 m [48]. Unless specified differently, due to a lack of available observations at the river
240 boundary the water temperature adapts to the environmental value inside the basin river
241 and the inflow salinity is fixed to a constant value of 0.1 psu. For the same reason, unless
242 specified differently, a constant concentration of 100,000 CFU 100 ml⁻¹ was imposed at
243 the source points during overflow events. Such value is in line with what reported in
244 the literature [7, and references therein].

245 The numerical model simulates the water circulation field, the water temperature,
246 and the salinity by representing the physical processes occurring in the coastal areas
247 of the Adriatic Sea, for example, tidal propagation, wind-induced currents and set
248 up, water, heat and salt fluxes, thermohaline stratification, and vertical mixing. The
249 simulations were performed for selected summer periods of 2019 and 2020.

250 In all cases, the numerical domain considers the area of interest and a larger part of
251 the coastal and shelf seas. To adequately resolve the river-sea continuum, the grids also
252 include the lower part of the considered river. The bathymetry interpolated onto the
253 numerical grids was obtained by merging high-resolution site-specific datasets covering
254 the area of interest with the composite EMODNet dataset [27] for the outer open sea. A
255 description of the different study areas, together with site-specific details of the model
256 implementation, is reported below.

257 2.2.1. Fano coast and Arzilla stream

258 The town of Fano is located along the Italian coast in the central Adriatic Sea and it
259 covers an area of 121 km² with 62,000 inhabitants and high urbanization. The coast of
260 Fano is low and sandy and characterized by several artificial protections against beach
261 erosion. The touristic harbour of Fano is located on the right side of the river mouth.
262 The Arzilla stream has a torrent-like character with water discharge ranging from less

263 than $1 \text{ m}^3 \text{ s}^{-1}$ up to $30 \text{ m}^3 \text{ s}^{-1}$. The Arzilla stream collects sewages from inland and
264 Fano combined sewer outflow and it discharges them into the sea, near one of the most
265 popular beaches during the summer season. Heavy rainfall often causes the overflow of
266 the local sewerage network that collects the microbial load from Fano town. Every time
267 that a sewer outflow occurs, the bathing activity in Fano is closed based on the potential
268 risk of faecal microbial contamination [49].

269 The resolution of the unstructured grid (8,675 triangular elements) ranges from
270 a few meters at the Arzilla stream mouth, up to 1.5 km at the open sea (Fig. 2). The
271 dimension of the finite elements covering the coastal area near the beaches is about
272 20 m. The water column is discretized in 15 vertical layers with variable thickness
273 ranging from 0.5 m, in the topmost 3 m, to 1 m in the deeper layers (the maximum
274 depth is 12 m). An hourly discharge time series obtained from water level measured
275 10 km upstream is prescribed at that river boundary and an estimated water discharge
276 of 50 l s^{-1} was imposed at the CSO when active (starting and ending times of sewer
277 outflow are constantly monitored). Observed *E. coli* concentration was applied at the
278 river open boundary.

279 2.2.2. Pescara coast and river

280 Pescara, located along the coast in the central Adriatic Sea, is the largest and, with
281 about 190,000 residents, the most populated urban settlement in the Abruzzo region.
282 The coast is low and sandy and the beach extends to both the north and south sides of
283 the Pescara River mouth. The coast is protected from beach erosion by emerged (in the
284 northern part) and submerged (in the southern part) artificial reefs. The river mouth is
285 delimited by hard structures consisting of a dike on its left side, the touristic harbour on
286 the right side and a breakwater located at approximately 400 m from the river end. The
287 average discharge of the Pescara River is $57 \text{ m}^3 \text{ s}^{-1}$. During heavy rainfall events that
288 exceed the capacity of the local sewerage systems, the Pescara river receives in its final
289 stretch (3.5 km from the mouth) wastewater through 8 principal CSOs.

290 The numerical grid consists of 14,394 triangular elements having resolution up
291 to 10 m and includes the Pescara River (up to 4 km upstream of the mouth where the
292 hydrological monitoring station is located) and a coastal area extending for about 4
293 km to the north and the south from the river (Fig. 2). All artificial coastal structures
294 were considered in the domain. In the present implementation, the model runs in the
295 zeta layer configuration, with 20 vertical layers of increasing thickness, from 0.5 m
296 in the topmost layers, up to 2 m in the deepest ones (the maximum depth is 18 m).
297 Freshwater discharge at a 15-minute frequency was available to force the model at the
298 river boundary. Measured or estimated wastewater discharge volumes were imposed at
299 the sewer outflows.

300 2.2.3. Raša River canal

301 The Raša River canal is a 14 km long and 700 m wide canal, with a depth ranging
302 from 1 m at the river mouth to 50 m at the sea boundary. This fjord-like environment
303 receives freshwater mainly from the Raša River, which has an average discharge of
304 about $5 \text{ m}^3 \text{ s}^{-1}$ and peak values exceeding $100 \text{ m}^3 \text{ s}^{-1}$ during flood events. Microbial
305 pollution originates mainly from the city of Labin which discharges partially treated
306 wastewaters in the Krapanj canal flowing into the Raša River near its mouth, and from
307 the Raša settlement which discharges untreated wastewaters at the mouth of the river.

308 The model application required a detailed bathymetric dataset interpolated on a
309 numerical mesh with horizontal resolution up to 10 m (Fig. 2). The numerical model
310 domain consists of 16,303 triangular elements and considers the lower part of the river,
311 the whole fjord and part of the shelf sea. However, the high horizontal resolution is not
312 enough to correctly describe hydrodynamics in this area. The vertical resolution and dis-
313 cretization become important when passing from really shallow and meandering zones,
314 like the inner river, to the shelf and, finally, to the open sea. In this model application,

we used a hybrid vertical coordinate system with 10 sigma layers in the upper 10 m and 2 m thick zeta layers in the deeper part. The choice of the hybrid system was driven by the need of resolving stratification even in the northern shallow part of the system where the river flows as well as of ensuring an adequate vertical resolution in the deeper part. Hourly water discharge values obtained from a monitoring station of Mutvica-Most located 6 km upstream of the river boundary were imposed in the numerical experiments. Water temperature and salinity continuously measured with 30-min frequency were used as boundary conditions of the Raša River. The estimated wastewater volume discharged in Labin and Raša for 2020 amounts to 322,024 and 150,000 m³ year⁻¹, respectively. Due to the lack of more detailed information, a continuous flow of wastewater was imposed at the boundaries.

2.2.4. Omiš coast and Cetina River

Omiš is a coastal town in the Dalmatia region of Croatia, located where the Cetina River meets the Adriatic Sea. The city has a population of about 15,000 inhabitants and is surrounded by sandy beaches and small pebble coves. The average discharge of the Cetina River is 18 m³ s⁻¹ with peak values reaching more than 300 m³ s⁻¹. The drainage system in the Omiš agglomeration is a combined drainage system consisting of a submarine outlet (1,600 m offshore at a depth of 60 m) discharging treated wastewater and 8 pumping stations (overflow elements) discharging untreated wastewaters in the Cetina River and along the coast during heavy rain events.

The numerical computation has been carried out on a spatial domain that represents part of the Cetina River (up to the discharge monitoring station of Tisne, 8 km upstream the mouth), the coastal area with the bathing sites and a portion of the sea limited southward by the island of Brač (Fig. 2). The unstructured grid is made of 6,406 triangular elements having a resolution ranging from 25 m close to the river mouth to 750 m in the open sea. The vertical discretization is based on a hybrid approach with 6 sigma layers in the topmost 10 m and 26 unevenly distributed zeta layers with thickness ranging from 2 to 5 m in the deeper open sea (the maximum depth of the grid is 72 m). An hourly discharge time series obtained from water level measured at Tisne was prescribed at that river boundary. The estimated wastewater volume discharged at the submarine outflow is 1,300,000 m³ year⁻¹ and the pumping stations have a capacity ranging from 10 to 150 l s⁻¹. Due to the lack of more detailed information, a continuous flow of wastewater was imposed at the boundaries.

2.2.5. Ploče coast and Neretva Estuary

The Neretva River flows near the port-town of Ploče in Croatia and represents one of the principal sources of freshwater in the Adriatic Sea with an average water discharge of about 300 m³ s⁻¹. This study area has around 35,000 inhabitants, and a wastewater system is partially established only in cities Ploče, Metković and Opuzen, but without treatment plants. In Ploče there are three outlets into the sea, two of which are located in the urban part of the city, and the third is located in the area of the Port of Ploče. In Opuzen and Metković, all the untreated water flows into the Neretva River, while wastewaters in Ploče are discharged into the sea.

Figure 2 reports the unstructured grid of the SHYFEM application, which considers the Neretva Estuary up to 20 km upstream the mouth (where the discharge monitoring station of Metković is located), nearby wetlands and part of the sea constrained by the coast and the Pelješac peninsula. The grid consists of 9,601 elements with a horizontal resolution varying from 50 m in the river and near the coast, to 750 m in the outer sea. Like many other coastal systems worldwide, the Neretva Estuary is subjected to the upstream extension of the mixing zone, with the consequent increase of the salt content in aquifers and surface waters [50]. To adequately account for the two-layer estuarine dynamics, the water column is represented by a hybrid vertical coordinate system consisting of 10 sigma layers in the upper 10 m and 18 zeta layers with a thickness

of 2 m (the maximum depth of the grid is 44 m). Hourly water discharges obtained from water levels observed in Metković were imposed at the river boundary. The estimated wastewater volume discharged in Metković, Opuzen and Ploče for 2020 amounts to 364,000, 70,000 and 210,000 m³ year⁻¹, respectively. Due to the lack of more detailed information, a continuous flow of wastewater was imposed at the point sources.

3. Results and Discussion

3.1. Evaluation of the modelling system

The applications of the SHYFEM model to the five study areas in the Adriatic Sea were validated by comparing various parameters. The model evaluation is limited by the availability of site-specific observations.

3.1.1. The observational datasets

The different study areas are monitored by several observational networks which differ for the observed parameters, type of monitoring instruments and frequency of acquisition. The monitored parameters used in the validation procedures are grouped into the following three categories:

- hydrodynamic: water levels;
- physicochemical: water temperature and salinity;
- microbial: faecal bacteria (*E. coli* and intestinal enterococci) concentration.

The main characteristics of the available dataset in the five study areas are presented in Table 1.

Table 1. Description of the observational datasets used for validating the modelling suite.

Study area	Hydrodynamic	Physicochemical	Microbial
Fano coast and Arzilla stream	None	Mid-column water temperature and salinity from water samples collected at the river mouth and along three coastal transects with points at 50, 100, 150, 200 and 250 m from the coastline. Nine monitoring surveys were performed in the summer of 2019 and 2020.	Mid-column <i>E. coli</i> concentration from water samples collected at the river mouth and along three transects with points at 50, 100, 150, 200 and 250 m from the coastline. Nine monitoring surveys were performed in the summer of 2019 and 2020.
Pescara coast and Pescara River	Water levels measured in the Pescara harbour at a 15-min frequency (2020)	Surface water temperature values measured in the Pescara harbour at a 15-min frequency (2020)	None
Raša River canal	None	Surface water temperature and salinity from water samples collected at the river mouth and along three transects with points at 200, 400 and 600 m from the river mouth, and at two popular touristic sites located at 1.5 and 3.4 km from the river mouth. Four monitoring surveys were performed in October and November 2020.	Surface <i>E. coli</i> concentration from water samples collected at the river mouth and along three transects with points at 200, 400 and 600 m from the river mouth, and at two popular touristic sites located at 1.5 and 3.4 km from the river mouth. Four monitoring surveys were performed in October and November 2020.
Omiš coast and Cetina River	Hourly water levels from Omiš, 1.4 km upstream of the river mouth (2020).	None	None
Ploče coast and Neretva Estuary	Hourly water levels measured in the Neretva Estuary at Opuzen, about 12 km upstream of the river mouth, and at Ušće, along the coast at 2.4 km from the river mouth (2020).	None	None

387 The monitoring stations in Pescara, Omiš-Cetina and Ploče-Neretva study areas
 388 are indicated with red dots in Fig. 2. The location of the physicochemical and microbial
 389 monitoring stations in the Fano-Arzilla and Raša River canal sites are shown in Fig. 9a
 390 and b, respectively.

391 3.1.2. Model assessment

392 The model performance was evaluated in terms of the difference between the
 393 average of simulated and observed values (BIAS), the root mean squared error (RMSE)
 394 and the Pearson product-moment correlation coefficient (R). For the concentration of
 395 *E. coli*, the root mean squared logarithmic (base 10) error (RMSLE) was used instead
 396 of RMSE [7]. Following the subdivision proposed for the observations, the model
 397 evaluation is presented firstly for hydrodynamics, afterwards for the physicochemical
 398 characteristics of the coastal waters and lastly for the microbial pollution.

399 *Hydrodynamic assessment*

400 Concerning the hydrodynamic assessment, the model results were compared with
 401 water levels recorded in Pescara, Omiš-Cetina and Ploče-Neretva study areas. The water
 402 level is here used to evaluate the hydrodynamic model performance. Observed and
 403 simulated time series were processed with a tidal harmonic analysis tool based on the
 404 least-squares fitting [51] to separate the tidal and the residual contributions to the total
 405 sea level. The statistics of the simulated values (total and tidal water levels) for the three
 406 study sites are reported in Table 2.

Table 2. Description of the observational datasets used for validating the modelling suite.

Study area	Station name	RMSE (m)	BIAS (m)	R
Pescara	Pescara harbour	0.17 / 0.02	0.02 / 0.	0.81 / 0.99
Omiš-Cetina	Omiš	0.09 / 0.03	0.04 / 0.	0.72 / 0.96
Ploče-Neretva	Ušće	0.08 / 0.02	-0.07 / 0.	0.79 / 0.98
	Opuzen	0.09 / 0.02	-0.07 / 0.	0.77 / 0.98

407 The model well reproduced the water levels variability observed in Pescara (top
 408 panel in Fig. 4), even if it is not able to capture the very high-frequency fluctuations,
 409 probably generated inside the harbour by resonance phenomena. RMSE, BIAS and R
 410 for the total water level are 0.17 m, 0.02 m and 0.81, respectively. However, the model
 411 simulated the tidal fluctuation (bottom panel in Fig. 4), which is the main driver of the
 412 sea-level variability in this area, with very high accuracy (RMSE=0.02 m and R=0.99).

413 The results of the model application to the Omiš-Cetina were compared with the
 414 water level continuously measured near the city of Omiš. The statistical parameters
 415 reported in Table 2 demonstrate that the model captures the sea-level variability in the
 416 investigated area, which was mostly determined by the tidal action. RMSE and R are
 417 0.09 and 0.72 for the total water level and 0.03 and 0.96 for the tidal level.

418 The numerical model well reproduced the water level also in the Ploče-Neretva
 419 study area (Table2) with an RMSE is 0.08 and 0.02 m for the total water level and the tidal
 420 level, respectively. The results of the tidal harmonic analysis revealed that the model
 421 captures the observed tidal amplification along the river estuary, even if it is slightly
 422 overestimating the amplitude of the K1 diurnal constituent. Generally, the comparison
 423 with the tide gauge data confirmed the good performance of the SHYFEM model in
 424 simulating sea levels and tidal propagation in the Adriatic Sea [34,48].

425 *Physicochemical assessment*

426 The water temperature and salinity values observed in the Fano-Arzilla, Pescara
 427 and Raša study areas were used to assess the capacity of the modelling system in
 428 reproducing heat fluxes, transport dynamics and mixing processes. Fig.5 shows scatter

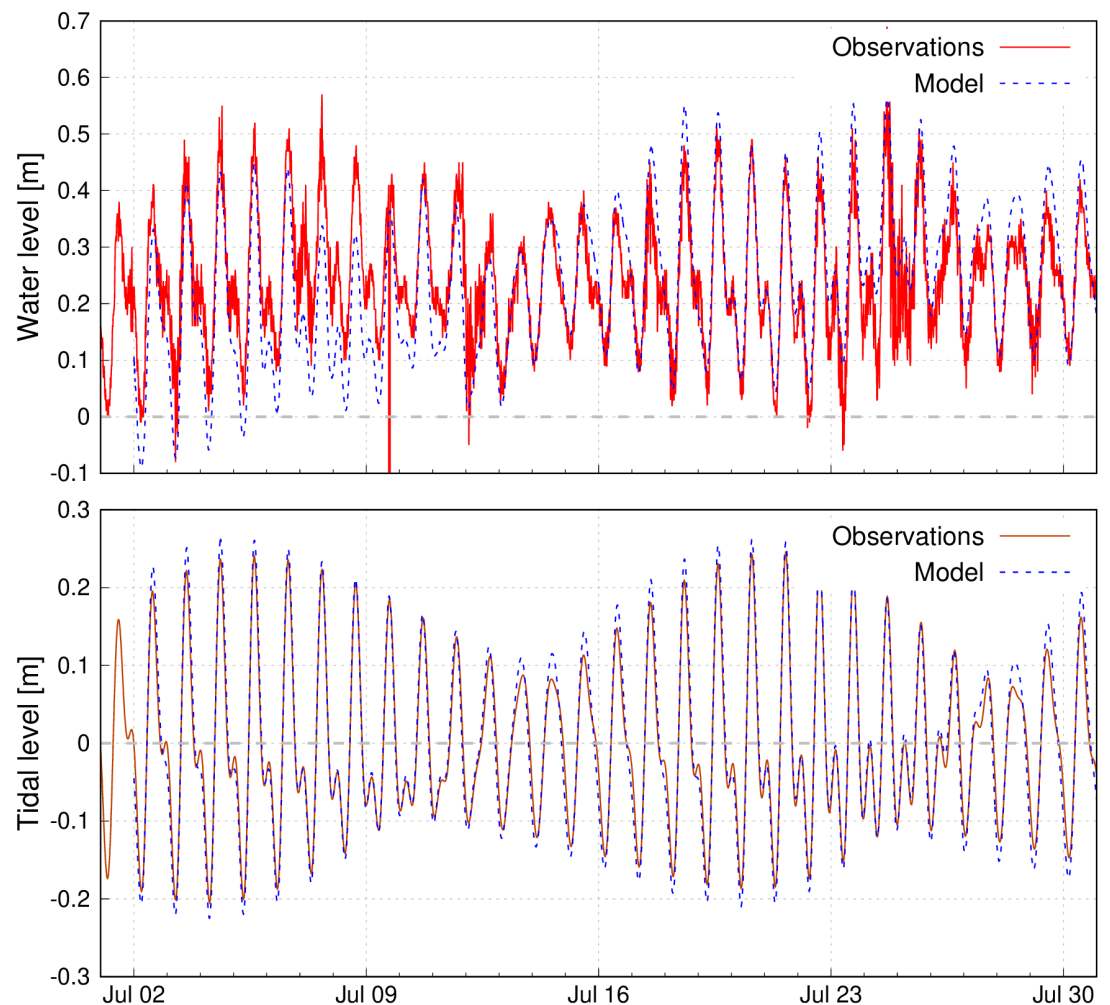


Figure 4. Measured and observed water levels in the Pescara harbour (summer 2020). The top panel presents the total water levels, while the bottom panel reports the tidal levels.

429 plots of simulated and observed water temperature (panel a) and salinity (panel b) for
 430 the Fano-Arzilla study area. The obtained BIAS and RMSE for salinity are 3.1 and 2.5
 431 psu, and $-0.1\text{ }^{\circ}\text{C}$ and $1.2\text{ }^{\circ}\text{C}$ for water temperature. The correlation coefficient resulted
 432 to be 0.95 and 0.64 for salinity and water temperature, respectively. The analysis of
 433 the results reveals that, despite the large uncertainty on the boundary conditions, the
 434 numerical model compares reasonably well with the measurements acquired in Fano
 435 coastal waters and reproduces the observed spatial and temporal variability of both
 436 water temperature and salinity. The model slightly overestimated salinity.

437 As shown in Fig.6, model results were generally in good agreement with the
 438 continuous water temperature values measured in the Pescara harbour. The model well
 439 captured the observed weekly variability of the water temperature during the summer
 440 of 2020, as well as the daily cycle. RMSE, BIAS and R between modelled and observed
 441 water temperatures in Pescara are $0.50\text{ }^{\circ}\text{C}$, $0.46\text{ }^{\circ}\text{C}$ and 0.93, demonstrating the good
 442 performance of the finite element modelling suite for this study site.

443 Despite the sparse data and the complexity of the system, the model seems to be
 444 able to reproduce the observed salinity and water temperature distributions in the Raša
 445 River canal (Fig.7). Salinity ranged from 3 to 38 psu and was generally increasingly
 446 moving from the river mouth to the sea, even if during the 18 September 2020 survey all
 447 observations have values around 37 psu. This is due to the temporal fluctuation of the
 448 Raša River discharge which in a few days passed from less than $1\text{ m}^3\text{ s}^{-1}$ to $15\text{ m}^3\text{ s}^{-1}$ as

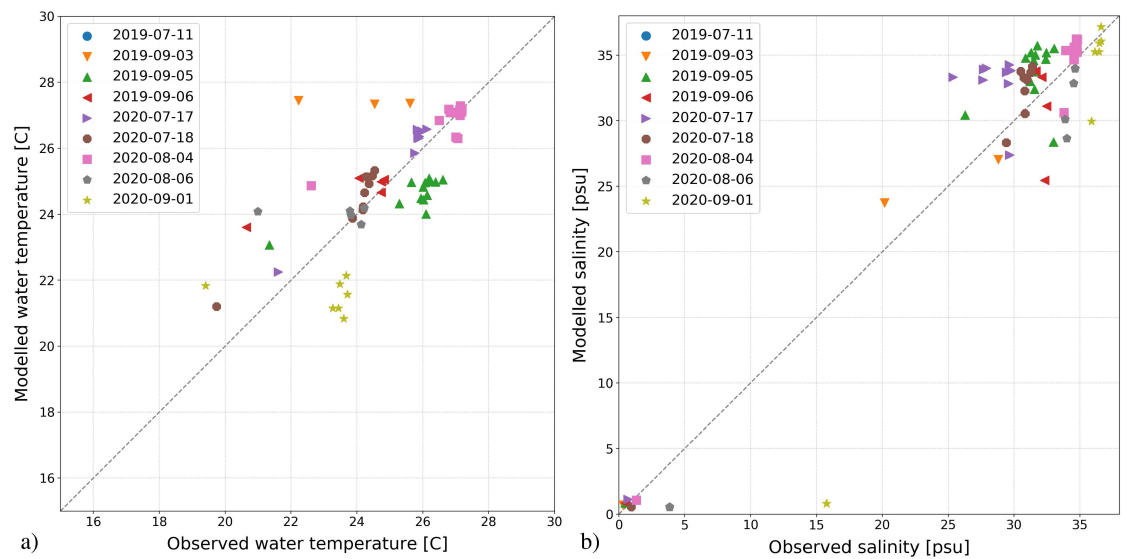


Figure 5. Scatter plot of observed and simulated water temperature (a) and salinity (b) in the Fano-Arzilla study area (2019 and 2020 samples).

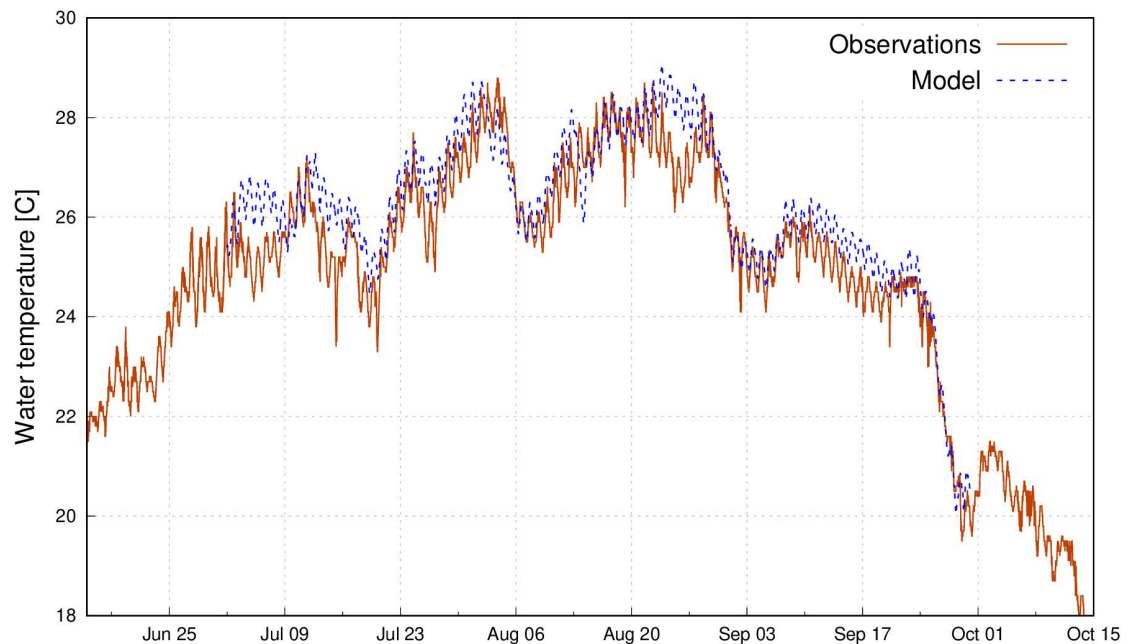


Figure 6. Measured and simulated water temperature in Pescara harbour (summer 2020).

449 a consequence of an intense rainy event. The obtained BIAS, RMSE and R for the salinity
 450 are 1.3 psu, 7.1 psu and 0.71, and -0.4 °C, 1.7 °C and 0.67 for the water temperature.
 451 Generally, the model underestimated salinity near the river mouth and overestimated it
 452 at the two touristic sites located at 1.5 and 3.4 km from the river mouth. The mismatch
 453 could be due to the uncertainty on the bathymetry of the very shallow (less than 1 m)
 454 area in front of the river mouth and which was not monitored during the bathymetric
 455 survey.

456 *Microbial pollution assessment*

457 Regarding microbial pollution, the numerical model results were compared with
 458 the *E. coli* concentration measured in the Fano-Arzilla and Raša study areas for assessing
 459 the capacity of the model in reproducing the dispersion and decay of faecal bacteria in
 460 nearshore waters. *E. coli* concentration is reported as CFU 100 ml^{-1} of water.

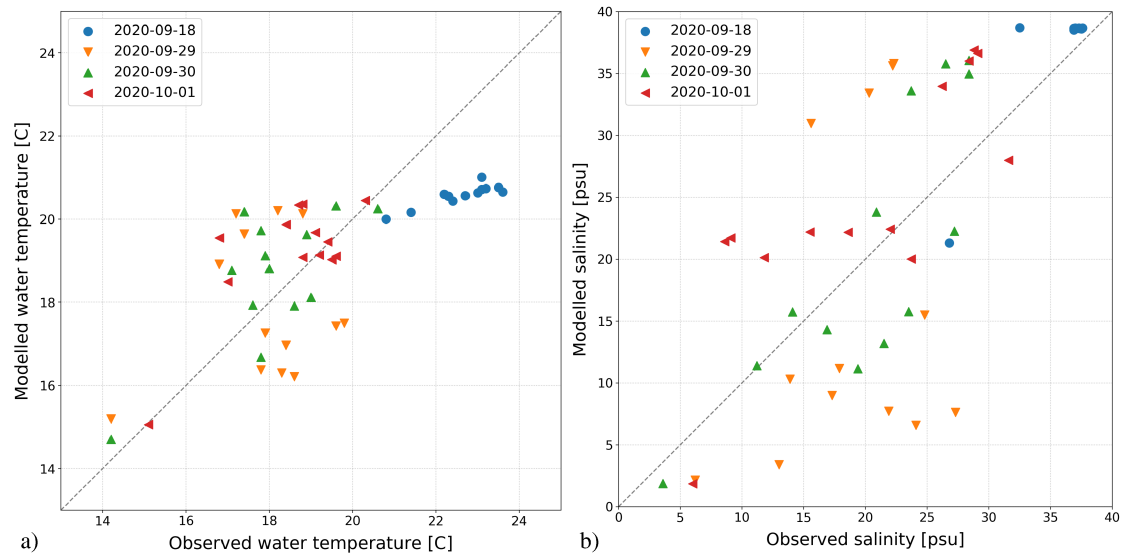


Figure 7. Scatter plot of observed and simulated water temperature (a) and salinity (b) in the Raša study area (2020 samples).

461 In the Fano-Arzilla site, the *E. coli* concentration was monitored with nine sampling
 462 surveys in the summer of 2019 and 2020. More details about the sampling strategy and the microbial analysis can be found in [49]. As shown in Fig. 8a, the numerical
 463 model provides a realistic representation of the *E. coli* distribution in the nearshore
 464 waters, describing the marked decrease in the bacteria concentration observed from
 465 the river mouth towards the open sea. This is mostly due to the effect of dilution with
 466 sea waters and decay induced by solar radiation and salinity. According to the scatter
 467 plot presented in Fig. 8b, the modelling system well described (mostly within an order
 468 of magnitude precision) the observed *E. coli* concentration measured in the two years
 469 of sampling activity. RMSLE for *E. coli* concentration in Fano-Arzilla is 0.18, a value
 470 below the ones reported in other studies [7,11,13,52], and the correlation coefficient
 471 is 0.93. During some events, e.g. on 5 September 2019 and 17 July 2020, the model
 472 underestimated the bacterial concentration in coastal waters. Such discrepancy could be
 473 related to the occasional formation of ephemeral stagnant freshwater pools at the river
 474 mouth, not reproduced by the model, where bacteria proliferate before reaching the sea.
 475

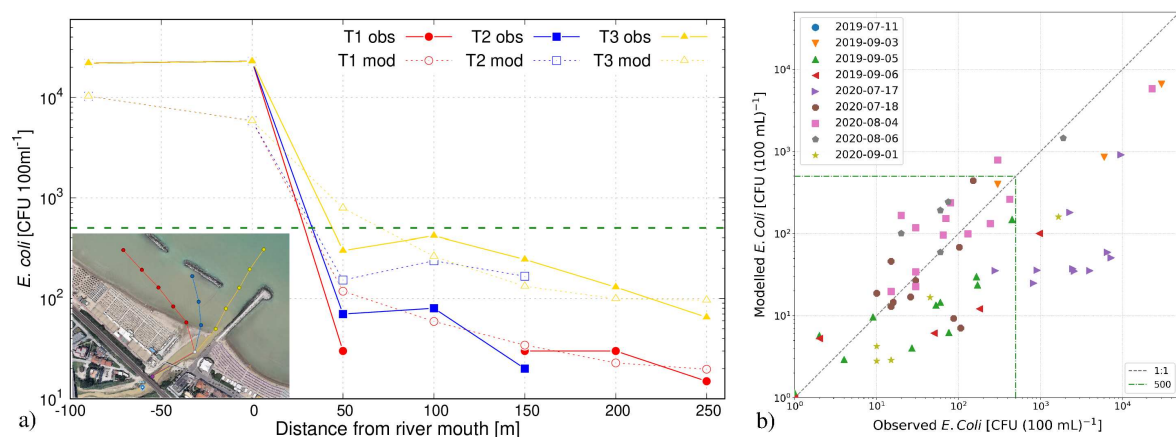


Figure 8. Observed (obs) and simulated (mod) *E. coli* concentration in Fano. a) Simulated (dashed lines) vs. observed (solid lines) concentration along the three river-sea transects monitored on 4 August 2020. b) Scatter plot of simulated versus observed values (2019 and 2020 samples). The green dashed line indicates the 500 CFU 100 ml⁻¹ value.

476 In the Raša River canal, the model is reproducing the observed *E. coli* concentrations
 477 with a satisfactory agreement (Fig. 9a). RMSLE and R for *E. coli* concentration in Raša
 478 are 0.44 and 0.68, respectively. *E. coli* concentrations at the mouth of the Raša River
 479 and adjacent touristic locations were below 10 CFU 100 ml⁻¹ on 18 September 2020
 480 and increased up to the bathing limit of 500 CFU 100 ml⁻¹ as a consequence of the
 481 rainfall rain event of 29 September 2020. As shown in Fig. 9b, the polluted waters
 482 coming from the Raša River tended to flow along the western coast. The model slightly
 483 underestimated the faecal bacterial concentration in Trget and Blaz.

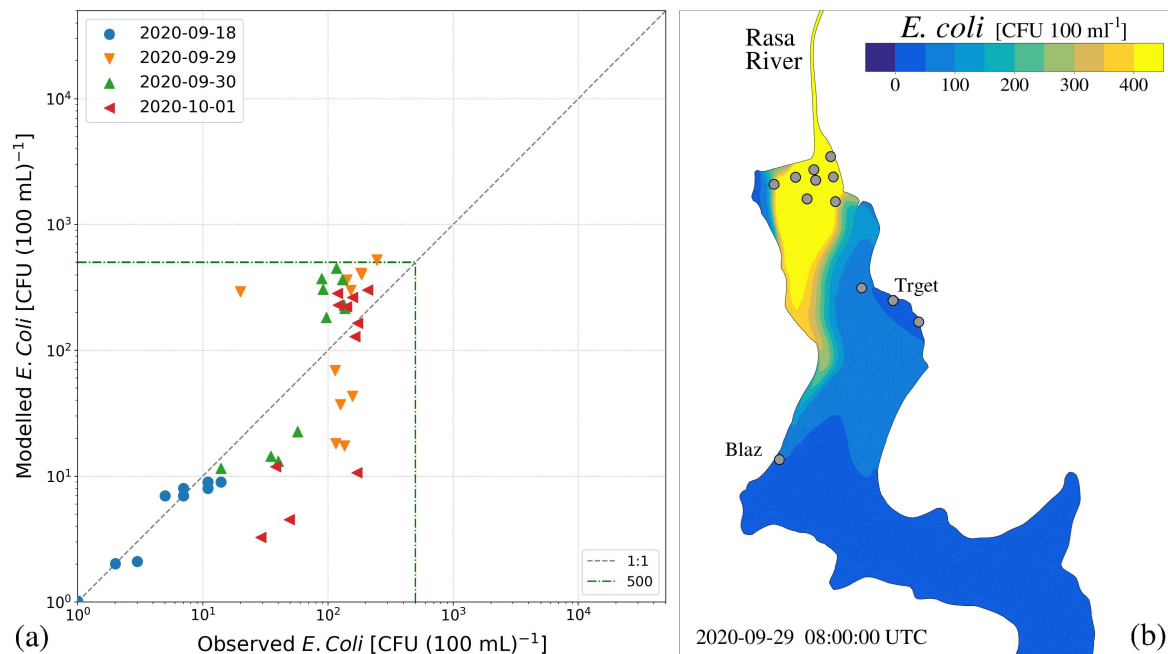


Figure 9. *E. coli* validation in Raša. a) Scatter plot of simulated versus observed *E. coli* concentrations (2020 surveys). b) Simulated distribution of *E. coli* on 29 September 2020 at 08 UTC. The grey dots mark the sampling stations.

484 3.2. Comparative analysis of the Adriatic study areas

485 A comparison study between the five study sites was carried out using the numerical
 486 model results. The analysis focused on the hydrodynamic characteristics and the
 487 quality of the bathing waters. Since this study concerned the contamination of
 488 recreational waters, the comparative analysis is based on the model results obtained
 489 for the summer of 2020, considered here as a common period of investigation for all
 490 study areas. We selected the summer months because they represent the period in which
 491 the bathing sites are mostly populated and microbial pollution events have the highest
 492 impact.

493 3.2.1. Circulation dynamics

494 In this section, we present and compare the hydrodynamic characteristics of the
 495 five study areas in terms of current and salinity patterns. The areas of investigation
 496 strongly differ for hydraulic and morphological characteristics. They are all coastal
 497 areas influenced by freshwater input, which, however, greatly differ for the discharged
 498 volumes and seasonal fluctuations. According to the morphological characteristics,
 499 we can classify the investigated area in sandy and mild sloping beaches with artificial
 500 barriers (Fano and Pescara coasts), gravel/rocky and steep shores (Omiš and Ploče
 501 coasts) and semi-enclosed coastal environments (Raša River canal). As a result of such
 502 variability and other forcing factors (the main characteristics of which are reported in
 503 Table 3), the oceanographic conditions of the study sites are driven by different dominant

504 processes which may determine different transport and diffusion dynamics. Salinity can
 505 be used as a proxy for the dispersion of contaminated water coming from the rivers.

Table 3. Characteristics of the main forcing factors (river flow, tidal range, sea temperature and wind speed) in the five study areas.

Study area	Tidal range [m]	River flow [$\text{m}^3 \text{s}^{-1}$] (mean/max)	Sea temperature [$^{\circ}\text{C}$] (mean/max)	Wind speed [m s^{-1}] (mean/max)
Fano-Arzilla	55	0.2 / 3.7	27 / 32	2.3 / 12.8
Pescara	34	38.0 / 60.0	27 / 29	3.8 / 12.2
Raša River	55	0.8 / 9.0	23 / 27	3.1 / 15.0
Omiš-Cetina	37	8.0 / 9.0	23 / 27	3.1 / 17.7
Ploče-Neretva	39	200.0 / 450.0	23 / 26	3.5 / 16.3

506 As a first step, let's have a look at the main surface circulation and salinity patterns
 507 in the different sites (Fig. 10). Even if the analysis focused on summer months, when
 508 freshwater inputs were at a minimum, generally, the water circulation near the river
 509 mouth was mainly driven by the river flow and its interaction with the coastal currents.
 510 In the Fano-Arzilla (Fig. 10a) and Pescara (Fig. 10b) sites, the main circulation resulted
 511 to be strongly influenced by the artificial structures (reefs and breakwaters) with the
 512 consequent deflection of the riverine waters towards the touristic western beaches.
 513 Similarly, the waters coming from the Cetina River tended to be deflected westward by
 514 the artificial jetty on the left-hand side of the mouth, thus determining the spread of
 515 the freshwater plume along populated beaches (Fig. 10d). On the contrary, in the Ploče-
 516 Neretva site (Fig. 10e), the jetties at the river mouth forced the outflowing freshwater
 517 to separate from the coast, thus decreasing the probability that polluted waters were
 518 transported to the bathing sites. During stratified summer conditions, the circulation in
 519 the Raša River canal was characterized by a surficial layer of water with reduced salinity -
 520 due to the freshwater supply - moving towards the open sea, primarily along the eastern
 521 side of the fjord (Fig. 10c). The strongest surface currents were found in Pescara due to
 522 the rather high and constant amount of freshwater discharge by the Pescara River.

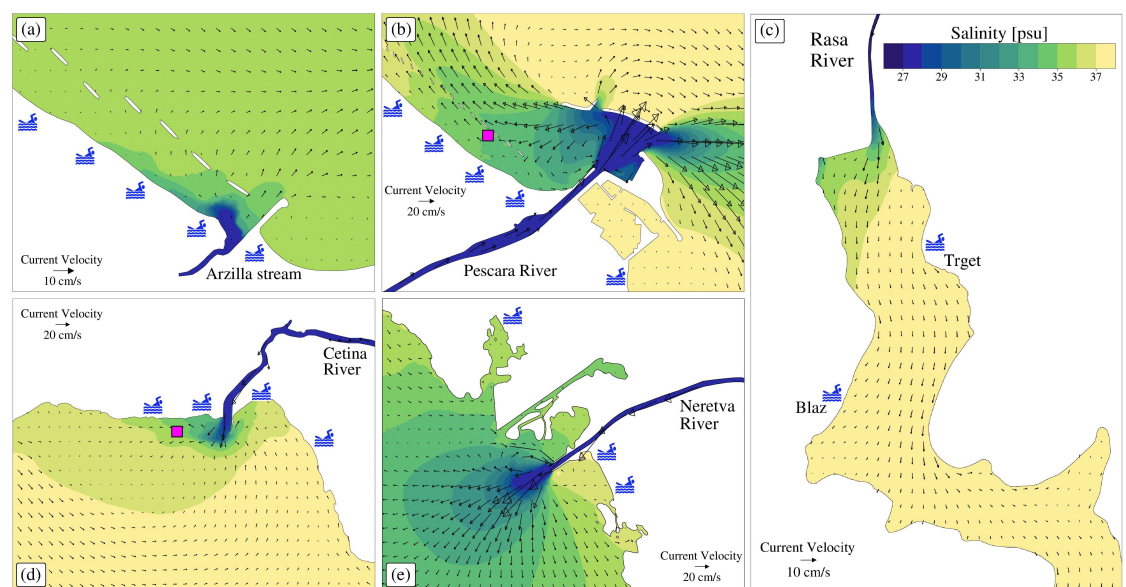


Figure 10. Mean surface current and salinity patterns in the five study areas (July-August 2020). The magenta squares mark the control stations presented in Fig. 11. The swimming symbols indicate the bathing locations.

523 Generally, the oceanographic conditions in the Adriatic Sea are characterized by
 524 stable thermal stratification in summer [25,53,54]. However, approaching the coast,

525 the vertical stratification of the water column could greatly differ in function of the
 526 morphological characteristics and forcing factors. From Simpson and Souza [55], we
 527 know that the short-term variability, due to tides and wind, interacts with baroclinic
 528 gradients producing vertical variations in the stability of the water column. Indeed,
 529 analysing the temporal evolution of the salinity and water temperature fields it emerged
 530 that the water column in the investigated coastal areas underwent periodic mixing. In
 531 Fig. 11, we present the timeseries of surface and bottom water temperature and salinity
 532 extracted from the model results in two study areas (Pescara and Omiš-Cetina) near a
 533 bathing site at a depth of about 2 m.

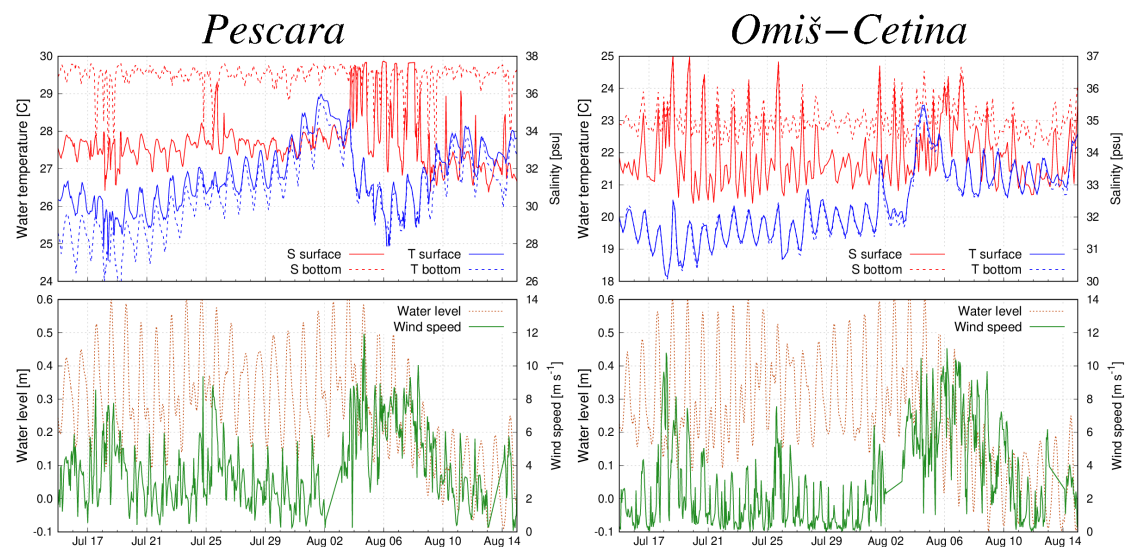


Figure 11. Top) Surface (solid line) and bottom (dotted line) water temperature (blue) and salinity (red) timeseries; Bottom) water level and wind speed timeseries. The values were extracted in the control stations in Pescara (left) and Omiš-Cetina (right) study areas (see the magenta squares in Fig. 10b and Fig. 10d, respectively).

534 Wind and tide concurred in mixing the water column. In Pescara (left panels in Fig.
 535 11), the relatively high amount of freshwater stratified the water column (with a 3 psu
 536 difference between surface and bottom salinity) and prevented mixing (when bottom
 537 water temperature and salinity values equal the surface ones), which occurred only
 538 during wind events with speed above 6 m s^{-1} (e.g. from 4 to 9 August 2020). In Pescara,
 539 water mixing was limited by the artificial reef, which confined part of freshwater masses
 540 near the beach. Stratification was also favoured by heat fluxes at the water surface. The
 541 tidal action modulated the water column stability by enhancing mixing during flood
 542 tide through the tidal straining mechanism [56]. Such an effect was more pronounced in
 543 the Omiš-Cetina site (right panels in Fig. 11), where the water column was fully mixed at
 544 a daily frequency. As already noted by [57], spring tides tended to produce well-mixed
 545 plumes while neap tides led to stratified plumes. Thermal stratification resulted to be
 546 more pronounced on the shallow western coast than the steep eastern shore. Similarly,
 547 coastal dynamics in the other study area (not shown) was regulated by wind, freshwater
 548 and tide.

549 Concluding, even if the Adriatic Sea is a micro-tidal environment, the tide is one
 550 of the main factors determining mixing in coastal areas [58]. Water mixing, and its
 551 variability in time and space, is crucial to be considered due to the dilution of polluted
 552 waters and the effect of temperature and salinity on faecal bacterial decay.

553 3.2.2. Quality of bathing waters

554 The hydrodynamic modelling presented in the previous section is devoted to the
 555 description of the water circulation under the influence of different forcing, but many
 556 substances are transported within the water. The concurrence of atmospheric forcing,

557 tide and freshwater inflows, led the Adriatic Sea to be characterized by a wide range of
 558 different transport phenomena. The analysis of the salinity patterns already provided
 559 indications on the transport processes, and also, on the coastal areas mainly influenced
 560 by river inputs. However, detailed numerical modelling of faecal bacteria was required
 561 for assessing the impact of microbial pollution on the quality of bathing waters.

562 Recreational waters in the investigated study areas were influenced by different
 563 sources of microbial pollution. The water quality in two Fano-Arzilla and Pescara study
 564 areas were influenced by urban sewage outfall triggered by heavy rainfall that exceeded
 565 the capacity of the sewerage systems of urban areas (<http://www.portaleacqua.salute.gov.it/Portale>).
 566 According to the regular monitoring activities, the three Croatian locations (Raša,
 567 Omiš-Cetina and Ploče-Neretva) had an excellent bathing water quality - on the basis of criteria
 568 defined by the EU bathing water directive - for the year 2020 (<http://baltazar.izor.hr/plazepub/ka>).
 569 Potential sources of microbial contamination of coastal waters are polluted river dis-
 570 charges and specific local discharges coming from legal and illegal sewer connectors.

571 To describe the transport, diffusion and decay of faecal bacteria in coastal waters,
 572 the performed simulations accounted for specific pollution events that occurred in
 573 the different study areas in summer 2020 and that were detected by the local bathing
 574 water management authority. The numerical results were processed to obtain maps of
 575 maximum *E. coli* concentration over the summer 2020 period (Fig. 12). Even if limited to
 576 a specific year of investigations, the maximum *E. coli* concentration maps provide clear
 577 indications of the zones more affected by microbial pollution.

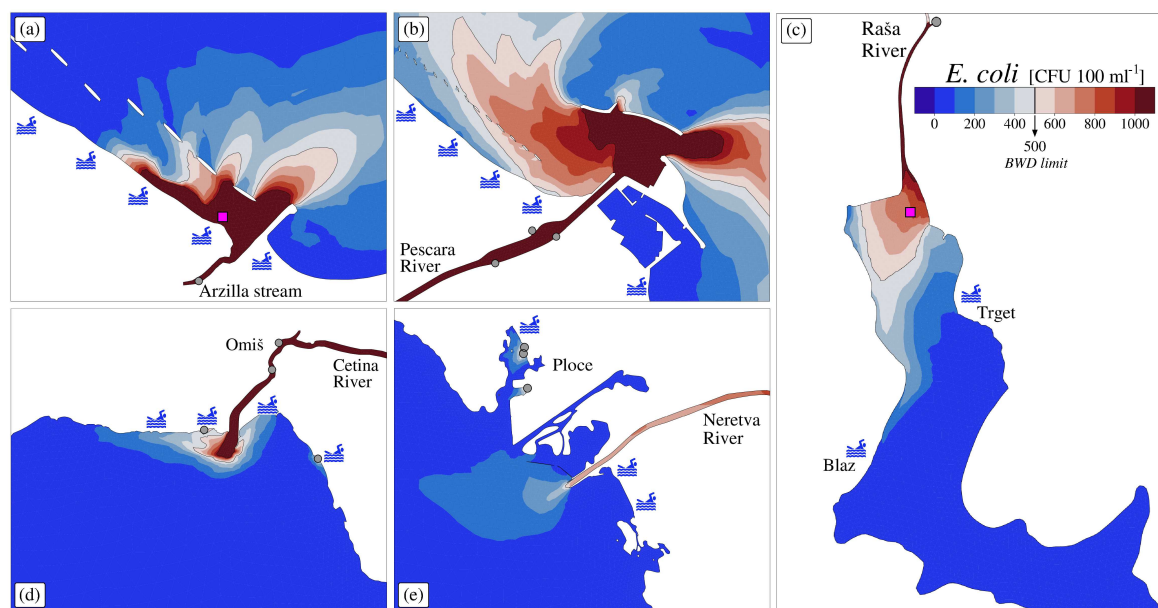


Figure 12. Maximum surface *E. coli* concentration maps in the five study areas (July-August 2020). The grey dots mark the point sources and the magenta squares mark the control stations presented in Fig. 13. The swimming symbols indicate the bathing locations.

578 In the Fano-Arzilla (Fig. 12a) and Pescara (Fig. 12b) sites, the *E. coli* plume extends
 579 from the river mouth and is constrained by the breakwater and artificial reefs which
 580 direct the flow of polluted waters towards the beaches. As a consequence, the nearshore
 581 *E. coli* concentration exceeded the BWD threshold values of 500 CFU 100 ml⁻¹. These
 582 two cases represent examples of inadequate planning of coastal defences, which were
 583 designed to protect beaches from erosion but have determined a worsening of bathing
 584 waters quality. In the Croatian pilot areas, the modelling results revealed for summer
 585 2020 a good water quality at almost all bathing locations of the Croatian study areas, as
 586 detected with the monitoring activity. In the Raša River site, *E. coli* concentration above
 587 the BWD threshold can be identified only near the river mouth and the polluted waters
 588 did not reach the bathing locations of Trget and Blaz (Fig. 12c). Similarly, in the bathing

589 locations of the Omiš-Cetina site (Fig. 12d), the *E. coli* concentration remained below
 590 the BWD limit due to the strong dilution of the polluted waters near the river mouth.
 591 Figure 12e shows the maximum *E. coli* concentration in the Ploče-Neretva site with the
 592 highest values found in the bay near the city of Ploče due to the discharge of wastewater
 593 from local point sources. The untreated waters discharged into the Neretva River at
 594 Opuzen and Metković are diluted by the freshwater flow and, consequently, the *E. coli*
 595 contamination from riverine waters had no significant impact on the coastal bathing
 596 sites. It has to be noted that only mean values of wastewater discharge were available
 597 for the Croatian sites and therefore the real bacterial concentrations can be higher than
 598 the simulated ones during intense pollution events.

599 In addition to the detailed spatial representation of microbial contamination, the
 600 numerical model allows for describing the temporal evolution of *E. coli* concentration
 601 during and after a pollution event. As an example, we report in Fig. 13 the timeseries
 602 of the *E. coli* concentration at a control station in the Fano-Arzilla site (indicated by the
 603 magenta square in Fig. 12a; depth of about 1 m) in August 2020 when two heavy rain
 604 events (Fig. 12b) in succession triggered sewer outflows. *E. coli* concentration in the
 605 coastal waters rose suddenly after the opening of the Arzilla spillway reaching a peak
 606 value of 10^4 CFU 100 ml^{-1} , well above the BWD threshold. The concentration remained
 607 above the threshold for about 12 hours and then decreased to values of about 100 CFU
 608 100 ml^{-1} . The analysis of the timeseries clearly shows that the concentration is strongly
 609 modulated by the tidal action (blue line in Fig. 12b) with peak values occurring during
 610 low tide. Concentrations above the threshold were found for about 3 days after the first
 611 rainy event. However, such peaks lasted for only a few hours per day.

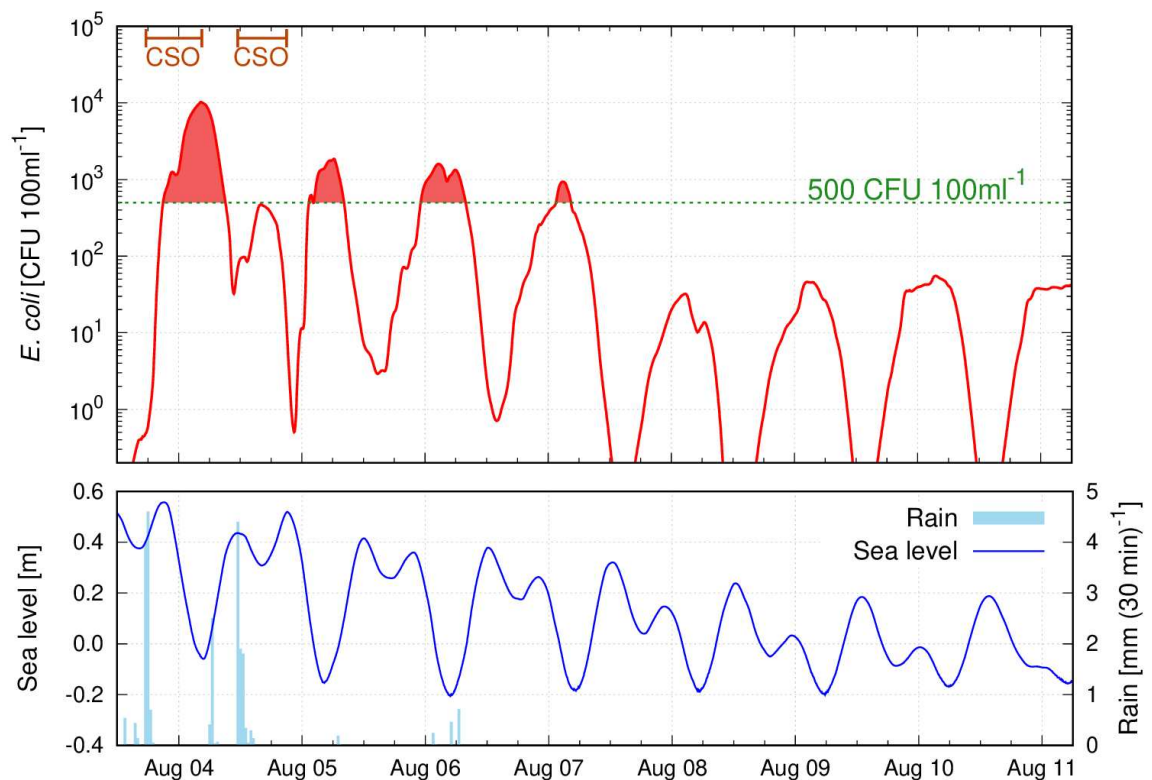


Figure 13. Time evolution of the modelled *E. coli* concentration in the Fano-Arzilla site (at the control station marked with the magenta square in Fig. 12a). The bottom panel reports rain and sea level.

612 As illustrated in Fig. 14, *E. coli* concentration in Raša River promptly responded to
 613 a river flood event and then was modulated by the tide, which determined a marked
 614 daily oscillation with values varying from 0 to $100\text{ CFU } 100\text{ ml}^{-1}$. The concentration

615 was extracted at a control station located at 400 m from the river mouth and having a
 616 mean depth of 2.7 m (magenta square in Fig. 12c). Even in such a shallow environment,
 617 an estuarine circulation system was present with the polluted riverine waters flowing
 618 on the surface. As a result, a strong vertical bacterial concentration is detected, with a
 619 two-order of magnitude difference in the bacterial concentration between surface and
 620 bottom. Similar results are found in the other investigated study sites, even the one
 621 experiencing a small tidal oscillation (not shown for lack of space).

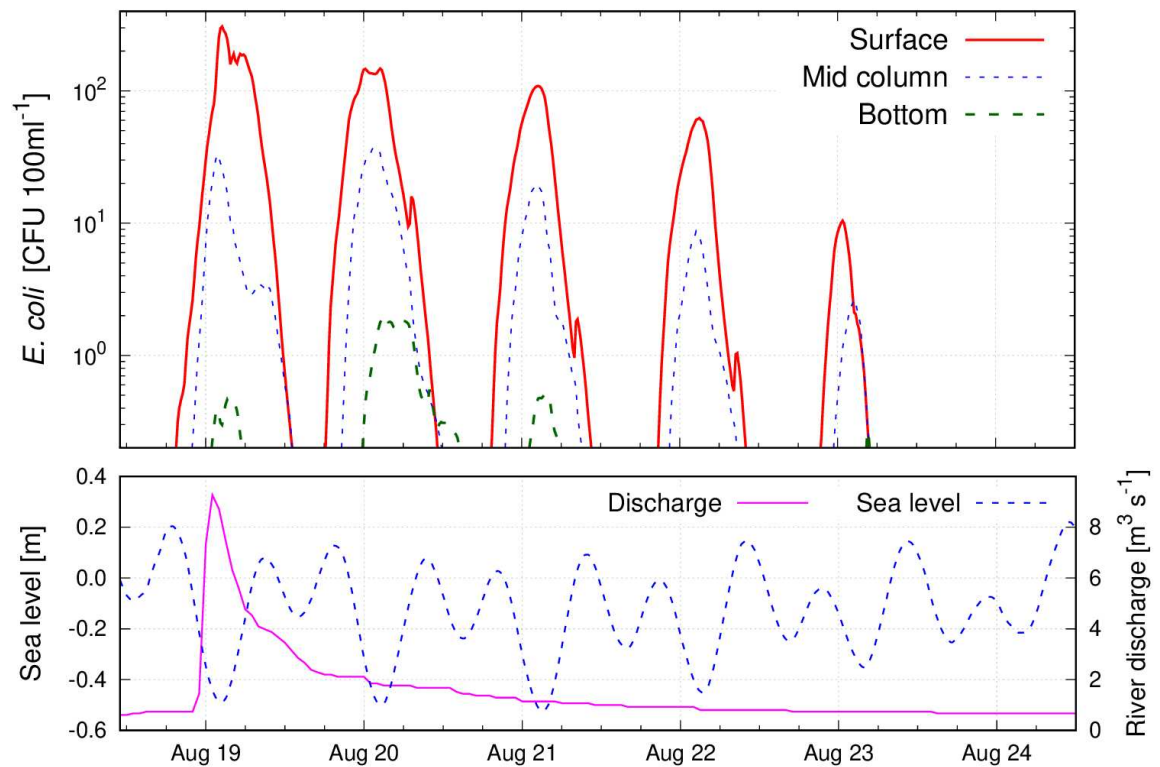


Figure 14. Time evolution of the modelled *E. coli* concentration in the Raša River site (at the control station marked with the magenta square in Fig. 12c). The bottom panel reports river discharge and sea level.

622 A crucial point in each environmental modelling application is the need for detailed
 623 forcing data. As demonstrated in this study, such a problem is particularly relevant for
 624 the modelling of faecal microbial contamination where detailed information on input
 625 sources from observations are mandatory for a realistic representation of the bacteria
 626 plume in coastal waters. However, the continuous monitoring of bacteria concentration
 627 in coastal seas is challenging and a large number of observational sites are required to
 628 correctly describe the interactions at the land-sea transition. This is especially true in
 629 coastal systems, as the ones investigated in this study, that are characterized by complex
 630 small-scale and high-frequency dynamics. The high horizontal, vertical and temporal
 631 variability of microbial contamination simulated by the model could not be detected
 632 by the ordinary monitoring activity that is performed at scheduled intervals in few
 633 stations (e.g. in the Fano-Arzilla site, 3 mid-column sampling stations every 15 days
 634 during the bathing season). Being the monitoring activity required to investigate the
 635 quality of bathing waters very expensive, numerical models - as the one presented in this
 636 study - can be very helpful for designing or optimizing monitoring networks [e.g. 59]
 637 Observations can also be assimilated into the model, increasing its capacity to represent
 638 the dynamics of the investigated system [60].

639 4. Conclusions

640 This study presents a relocatable coastal water quality prediction model - consisting
641 of hydrodynamic, dispersion and decay modules - that can be used for investigating
642 the spatial and temporal evolution of microbial pollution in coastal ecosystems. The
643 developed water quality model was successfully applied and validated in several coastal
644 areas facing the Adriatic Sea. Numerical model results demonstrated that, in the Adriatic
645 Sea, dilution and mixing had a stronger effect on bacteria reduction with respect to
646 microbial decay induced by base mortality, water temperature, salinity and sunlight.
647 Generally, the estuarine circulation near the river mouth favoured the seaward transport
648 of polluted riverine waters during the decreasing tide and obstructed the river outflow
649 during the rising tide. Due to the thermohaline stratification, strong vertical gradients of
650 bacterial concentration were found at the considered bathing sites.

651 The comparative analysis among the different study sites revealed a high spatial
652 and temporal variability of the circulation and dispersion dynamics in coastal waters,
653 which cannot be adequately described by the monitoring activity. Therefore, even if
654 each numerical model is a partial, simplified and mostly inaccurate representation of the
655 real world, it can be used for complementing the collected information retrieved by the
656 direct microbial monitoring. The synergic use of in situ observations and models allows
657 a reduction of uncertainties in studying coastal waters and improves our knowledge of
658 those regions also leading to further improvements in developing microbial monitoring
659 and modelling techniques.

660 With this perspective in mind, the numerical model described in this study was
661 developed as part of the Water Quality Integrated System (WQIS) proposed in the
662 WATERCARE project (<https://www.italy-croatia.eu/web/watercare>), an EU Interreg
663 Italy-Croatia project with the objective of reducing the impact of microbial environment
664 contamination in Adriatic bathing waters. WQIS is composed of a real-time hydro-
665 meteorological monitoring network, an automatic refrigerated sampling system, a water
666 quality monitoring network and a forecast operational modelling suite [49]. The model
667 applications described here will be made operational for providing bathing quality
668 forecasts with the aim of helping the management of faecal bacteria pollution in coastal
669 waters. The relocatable modelling suite presented in this study, as well as the whole
670 WQIS, can be easily implemented in other coastal systems.

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689 Abbreviations

- 690 The following abbreviations are used in this manuscript:
- | | |
|------------|---|
| BWD | EU Bathing Water Directive [8] |
| 691 SHYFEM | System of Hydrodynamic Finite Element Modules |
| WQIS | Water Quality Integrated System |

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