POLITECNICO DI TORINO Repository ISTITUZIONALE

Climate change impact on the ecological status of rivers: The case of Albaida Valley (SE Spain)

Original

Climate change impact on the ecological status of rivers: The case of Albaida Valley (SE Spain) / Vagheei, Hamed; Laini, Alex; Vezza, Paolo; Palau-Salvador, Guillermo; Boano, Fulvio. - In: SCIENCE OF THE TOTAL ENVIRONMENT. - ISSN 0048-9697. - 893:(2023). [10.1016/j.scitotenv.2023.164645]

Availability: This version is available at: 11583/2979432 since: 2023-07-04T10:09:57Z

Publisher: Elsevier

Published DOI:10.1016/j.scitotenv.2023.164645

Terms of use:

This article is made available under terms and conditions as specified in the corresponding bibliographic description in the repository

Publisher copyright

(Article begins on next page)

Contents lists available at ScienceDirect





Science of the Total Environment

journal homepage: www.elsevier.com/locate/scitotenv

Climate change impact on the ecological status of rivers: The case of Albaida Valley (SE Spain)



Hamed Vagheei^{a,*}, Alex Laini^b, Paolo Vezza^a, Guillermo Palau-Salvador^c, Fulvio Boano^a

^a Department of Environment, Land, and Infrastructure Engineering, Politecnico di Torino, Turin, Italy

^b Department of Life Sciences and Systems Biology, Università degli Studi di Torino, Turin, Italy

^c INGENIO CSIC-UPV, Universitat Politècnica de València, Valencia, Spain

HIGHLIGHTS

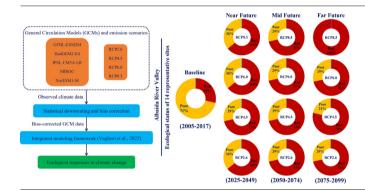
GRAPHICAL ABSTRACT

- A recently presented modeling framework is employed in different scenario simulations.
- Climate change impact on ecological status is investigated.
- Climate change influences hydrology, water quality, and ecological status.
- In certain representative sites, ecological status is expected to downgrade.
- Scientifically informed decisions are essential to manage and protect freshwaters.

ARTICLE INFO

Editor: Christian Herrera

Keywords: Ecohydrological model Freshwater Ecological status Biodiversity SWAT



ABSTRACT

Understanding the effects of environmental stressors (e.g., potential changes in climate and land use) on ecological status is essential for freshwater management. The ecological response of rivers to stressors can be evaluated by several physicochemical, biological, and hydromorphological elements as well as computer tools. In this study, an ecohydrological model based on SWAT (Soil and Water Assessment Tool) is used to investigate climate change impact on the ecological status of Albaida Valley Rivers. The predictions of five General Circulation Models (GCMs) each with four Representative Concentration Pathways (RCPs) are employed as input to the model for simulating several chemical and biological quality indicators (nitrate, ammonium, total phosphorus, and the IBMWP (Iberian Biological Monitoring Working Party) index) in three future periods (Near Future: 2025-2049, Mid Future: 2050-2074, and Far Future: 2075-2099). Based on chemical and biological status predicted with the model, the ecological status is determined at 14 representative sites. As a result of increased temperatures and decreased precipitations from most of GCMs projections, the model predicts decreased river discharge, increased concentrations of nutrients, and decreased values of IBMWP for future compared to the baseline period (2005-2017). While most representative sites have poor ecological status (10 sites with poor ecological status and four sites with bad ecological status) in the baseline, our model projects bad ecological status for most representative sites (four sites with poor ecological status and 10 sites with bad ecological status) under most emission scenarios in the future. It should be noted that the bad ecological status is projected for all 14 sites under the most extreme scenario (i.e., RCP8.5) in the Far Future. Despite the different emission scenarios, and all possible changes in water temperature and annual precipitation, our findings emphasize the urgent need for scientifically informed decisions to manage and preserve freshwaters.

1. Introduction

Corresponding author.
E-mail address: hamed.vagheei@polito.it (H. Vagheei).

A variety of ecosystem services including provisioning services (e.g., irrigation and power generation), regulatory services

http://dx.doi.org/10.1016/j.scitotenv.2023.164645

Received 6 March 2023; Received in revised form 1 June 2023; Accepted 1 June 2023

Available online 14 June 2023

0048-9697/© 2023 The Authors. Published by Elsevier B.V. This is an open access article under the CC BY license (http://creativecommons.org/licenses/by/4.0/).

(e.g., maintenance of water quality), cultural services (e.g., recreation and tourism), and supporting services (nutrient cycling and ecosystem resilience) are provided by freshwaters (Aylward et al., 2005). However, climate change, that is one of the most severe environmental issues, constitutes a menace to global freshwater sources (UN-Water, 2020). While past and current global warming has impacted freshwater resources and caused significant changes in their communities, the risks of biodiversity loss due to future climate change are expected to be even greater (IPCC, 2022).

Addressing and managing the effects and challenges posed by climate change on freshwater environments, requires a holistic and fundamental understanding of linkages between climate, hydrology, water quality, and biodiversity. These connections can be studied by integrating water quality monitoring and modeling (Holguin-Gonzalez et al., 2013; Vagheei et al., 2022). Water quality monitoring, that encompasses the collection of information on various biological, physico-chemical, and hydromorphological components of water resources (e.g., macrophytes, macroinvertebrates, nutrients, turbidity, and river discharge), is a useful tool for the detection of water pollution, the determination of ecological status, and the management and restoration of freshwater environments (Strobl and Robillard, 2008; Canter, 2018). In addition, the integrated modeling approaches can supplement the data obtained from field monitoring by producing reliable predictions of freshwater ecosystems responses to different climate scenarios (Guse et al., 2015; Carr et al., 2019).

Recent model-based studies predict that climate change can affect hydrological components (e.g., streamflow and evapotranspiration) (Goodarzi et al., 2020; Mahmoodi et al., 2021; Tarekegn et al., 2022), water quality variables (e.g., nutrient loads) (Povilaitis et al., 2018; Shrestha et al., 2018), and aquatic communities (e.g., abundance and diversity of fish and macroinvertebrates species) (Guse et al., 2015; Kakouei et al., 2018; Theodoropoulos and Karaouzas, 2021) of river watersheds. Climate change would likely exacerbate the frequency, duration, and severity of droughts as it has been reported in Africa by Ahmadalipour et al. (2019), in Aragon (NE Spain) by Gaitán et al. (2020), and in the Miyun Reservoir Watershed (China) by Qiu et al. (2023) that may be associated with the increased concentration of nutrients due to the reduction of water sources (Whitehead et al., 2006; Charlton et al., 2018). Climate change would also intensify extreme rainfall events as it has been reported over different climate regions by Tabari (2020) and across the U.S. by Moustakis et al. (2021) that may result in increased surface runoffs carrying more nutrients to water bodies (Andersen et al., 2006; Shrestha et al., 2012). Reduced river flow, higher level of nutrient concentration, excessive nutrient loading, and higher temperature caused by climate change then would impact freshwater habitat and biodiversity as it has been reported in the Tehuacán-Cuicatlán Biosphere Reserve (Mexico) by López-López et al. (2019) and in the Athabasca River Basin (Canada) by Morales-Marín et al. (2019).

Even though previous studies predict a detrimental effect of global warming on rivers (Jun et al., 2010; Lutz et al., 2019), the relationship between climate change and ecological status is still largely unexplored in many river watersheds. Therefore, this study uses the integrated modeling framework presented by Vagheei et al. (2022) to investigate possible impact of climate change on ecological conditions of the Albaida Valley Rivers (SE Spain), which, to the best of our knowledge, this valley has not previously been studied under possible future climate change scenarios. This study follows the Water Framework Directive (WFD) 2000/60/EC, that is the main legislation for achieving good ecological status in Europe, to specify the ecological status of the rivers. According to the WFD, the estimation of the overall ecological status is based on the status of biological quality elements (i.e., macroinvertebrates, diatoms, macrophytes, and fish) and supporting quality elements (i.e., physico-chemical and hydromorphological) (Tueros et al., 2009; Zacharias et al., 2020). The concept behind the WFD 2000/60/ EC is the "One Out-All Out" (OOAO) principle, that assumes the final ecological status to be the worst status of the quality elements employed in the assessment (Zacharias et al., 2020). In the present work, the future ecological status of 14 representative sites along the rivers of the valley is predicted based on long time continuous projections of several chemical and biological indicators including nitrate, ammonium, total phosphorus, and the IBMWP

(Iberian Biological Monitoring Working Party) index (a well-known macroinvertebrate-based metric) under several climate scenarios. Although the ecological status is generally assessed with more than one biological group, the macroinvertebrate community (the IBMWP index) is only used in this work as data on fish, diatoms and macrophytes were not available. However, macroinvertebrates are by far the most used and reliable biological group for biomonitoring, sometimes the only group that is used for river biomonitoring (Birk et al., 2012; Bo et al., 2017; Vitecek et al., 2021). In addition, the choice of this work is to isolate the effects of climate change on the Albaida Valley, and other factors such as demographic change, land use change, socioeconomic development, and changes in environmental strategies, which are important for optimal management of watersheds, are thereby not included. In fact, while the increased demand for water, energy, and food due to population and economic growth (Fader et al., 2016; Wang et al., 2019) leads to land use changes with respect to urbanization, industrialization, agriculture intensification, and deforestation (Aghsaei et al., 2020; Lei et al., 2022) and may increase the deterioration of freshwater ecosystems (Tzanakakis et al., 2020; Peña-Angulo et al., 2021), the aim of the present work is to understand how the processes induced by climate change can impact river water quality. The chosen approach in the present work allows quantifying and characterizing the uncertainties from both the model and the future climate projections, which has been discussed as a crucial step for watershed modeling and climate change impact assessments (Deser et al., 2012; Abbaspour et al., 2015). In fact, this study considers the model prediction uncertainty using the Sequential Uncertainty FItting version 2 (SUFI-2) algorithm and deals with uncertainties associated with future climate projections using ensemble modeling, that is the implementation of several climate models and emission scenarios.

2. Materials and methods

Climate change scenarios are simulated to assess the impact of changing future conditions on ecological status of the Albaida Valley (Spain). For this purpose: (1) Predictions of General Circulation Models (GCMs) are statistically downscaled, and bias corrected with the observed climate data; (2) the bias-corrected GCM data are coupled to the ecohydrological modeling framework proposed by Vagheei et al. (2022) to evaluate the ecological responses of Albaida Valley to climate change. Fig. 1 shows the overall methodology applied in the study, which is explained in detail in the next sections.

2.1. Study area and data description

A shrubland and agricultural dominated part of the Albaida Valley (Spain) with an approximate area of 320 km² is studied (Fig. 2). The Albaida Valley, which is characterized by a semi-arid Mediterranean climate, consists of the Clariano (P1-P9, Fig. 2) and Albaida Rivers (P10-P14, Fig. 2). The behavior of the Albaida River differs upstream and downstream of the confluence with the Clariano River (P13, Fig. 2) as it receives important contributions from the Clariano River that is severely affected by the effluents of WasteWater Treatment Plants (WWTPs). A few kilometers downstream of their confluence, the Albaida River flows into the Bellús Reservoir which is one of the most eutrophic reservoirs in the Valencian Community. A complete description of the area can be found in Vagheei et al. (2022).

In addition to data used for developing the ecohydrological model of the valley (Vagheei et al., 2022), the daily precipitation, and minimum and maximum temperature of five GCMs and four Representative Concentration Pathways (RCPs) with a resolution of 0.5° supported by ISI-MIP5 (Inter-Sectoral Impact Model Intercomparison Project) (Hempel et al., 2013) collected from https://www.2w2e.com and the daily observed precipitation, and minimum and maximum temperature data with a resolution of 5 km from the Spanish National Meteorological Service (AEMET) collected from https://swat.tamu.edu are also used (Table 1). The observed climatic data are used for downscaling and bias correction of the GCM results. Table S1 in the Supplement provides general information on the locations of the climatic data obtained from the monitoring stations and the GCMs.

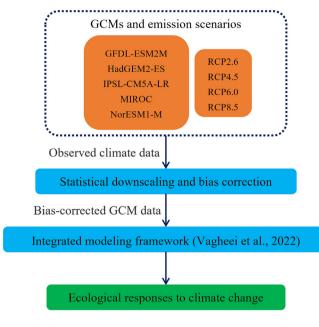


Fig. 1. Overview of the methodology applied in this study.

2.2. Uncertainty in climate change impact assessments

The assessment of climate change effects on river watersheds is accompanied with uncertainties from both the watershed model (e.g., uncertainties associated with the model input and the model structure) and climate projections (e.g., uncertainties associated with GCMs, emission scenarios, and downscaling and bias correction methods) (Kundzewicz et al., 2018; Tang et al., 2019). As previous studies found conflicting results about the relative importance of the uncertainty associated with climate projections compared to the hydrologic model uncertainty (Bennett et al., 2012; Sellami et al., 2016), characterizing uncertainty from both climate projections and the model is important for climate change impact assessments (Deser et al., 2012; Abbaspour et al., 2015). Thus, outputs of five GCMs including GFDL (US), HadGEM (UK), IPSL (FR), MIROC (JA), and NoerESM (NO) each with four carbon emission scenarios (RCP2.6, RCP4.5, RCP6.0, and RCP8.5) are used in the present work, as ensemble modeling is recommended in several studies to deal with uncertainties associated with climate projections (Semenov and Stratonovitch, 2010; Teutschbein and Seibert, 2012; Perez et al., 2014; Shen et al., 2018). In addition, the SUFI-2 algorithm is used to characterize the model prediction uncertainty as it is capable of implementing all model uncertainties on the parameters (Abbaspour et al., 2015). In SUFI-2, 500 model realizations are simulated and the propagation of the uncertainty in parameters leads to the model prediction uncertainty which is quantified by the 95 % prediction uncertainty (the 95PPU). The 95PPU is calculated from the 2.5 % and 97.5 % quantiles of cumulative distribution of output variables generated through Latin hypercube sampling (Abbaspour, 2015).

2.3. Preparation of future climate data

GCMs, which are three-dimensional numerical models that represent the physical dynamics of the atmosphere, ocean, cryosphere, and land surface, have been developed to project how these systems would behave under possible future greenhouse gas (GHG) and aerosol forcing (IPCC, 2013). Projections from GCMs, which are time series of climatic data such as precipitation and temperature, are the best available data that could be used as input to ecohydrological models to simulate the effects of climate change on water resources, nutrient loading, biodiversity, ecosystems, and, eventually, to improve watershed decision-making and

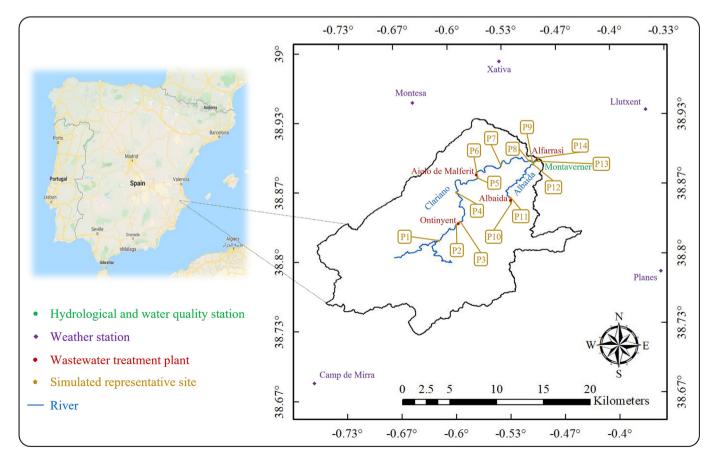


Fig. 2. Location of the study area, hydrological, water quality and meteorological stations, WWTPs, and representative sites simulated with SWAT (Soil and Water Assessment Tool).

Table 1

Overview of climate data used in this study. GFDL-ESM2M: Geophysical Fluid Dynamics Laboratory-Earth System Model version 2 M; HadGEM2-ES: Hadley Centre Global Environmental Model version 2-Earth System; IPSL-CM5A-LR: Institute Pierre Simon Laplace-Climate Model version 5A- Low Resolution; MIROC: Model for Interdisciplinary Research on Climate; NorESM1-M: Norwegian Earth System Model version 1 M (intermediate resolution).

Database	Model	Scenario	Period	Source	
GCM1	GFDL-ESM2M	RCP (2.6, 4.5, 6.0, 8.5)	1951-2099	NOAA/Geophysical Fluid Dynamics Laboratory	
GCM2	HadGEM2-ES	RCP (2.6, 4.5, 6.0, 8.5)	1951-2099	Met Office Hadley Center	
GCM3	IPSL-CM5A-LR	RCP (2.6, 4.5, 6.0, 8.5)	1951-2099	L'Institut Pierre-Simon Laplace	
GCM4	MIROC	RCP (2.6, 4.5, 6.0, 8.5)	1951-2099	AORI, NIES and JAMSTEC	
GCM5	NoerESM1-M	RCP (2.6, 4.5, 6.0, 8.5)	1951-2099	Norwegian Climate Center	
Observed data			1951-2005	AEMET (Spain)	

management (Bouraoui et al., 2002; Park et al., 2013; Guse et al., 2015; Marcinkowski et al., 2017). The GCM data need to be downscaled and bias-corrected before being used in regional impact analyses as their spatial resolutions are generally too coarse and, furthermore, all GCM projections contain some biases that can result in significant errors, if not corrected (Teutschbein and Seibert, 2012; IPCC, 2013). Thus, the Climate Change Toolkit (CCT) (Ashraf Vaghefi et al., 2017) is used in this study to downscale and bias-correct the data of five previously mentioned GCMs for the study region. The toolkit identifies the measured data stations that are closest to the GCMs stations to apply the correction factor: for temperature, the "additive correction method" is employed, and for precipitation, the "ratio correction method" is used (Ashraf Vaghefi et al., 2017).

2.4. Projection of ecological status

In the present work, the ecological status of 14 representative sites (P1-P14, Fig. 2) is studied based on a successfully implemented SWAT-based model of the Albaida Valley River Basin (Vagheei et al., 2022). According to the WFD, the determination of the overall ecological status involves considering both biological quality elements (macroinvertebrates, diatoms, macrophytes, and fish) and supporting quality elements, which are physico-chemical and hydromorphological (Tueros et al., 2009; Zacharias et al., 2020). However, hydromorphological status assessment for water bodies with an ecological status less than "high". Therefore, chemical and biological indicators are only used in this study. In addition, only the macroinvertebrate community is used among all biological indicators for assessing the biological status because of the lack of data on fish, diatoms and macrophytes.

To study the impact of climate change on river runoff and nutrient concentrations (concentrations of nitrate, ammonium, and total phosphorus), scenario runs based on the bias-corrected data of five GCMs under four carbon emission scenarios are compiled and analyzed with the SWAT model. The evaluation of ecosystem health in Near, Mid, and Far Future is then predicted with the IBMWP index (not only the most widely used index in Spanish Mediterranean rivers (Guareschi et al., 2017), but also the required index by the Spanish Royal Decree 817/2015), which was shown to be well correlated to nitrate concentrations (see Vagheei et al. (2022) for more details on this correlation). For this purpose, the SWAT model setup in the previous study (Vagheei et al., 2022) is updated as follows: (1) the Hargreaves method (Hargreaves et al., 1985) is used for estimating potential evapotranspiration instead of the Penman-Monteith method as GCM data consists only of precipitation and temperature data; (2) the land use map and agricultural management operations such as fertilization and irrigation are assumed not to change, and the average monthly effluent flow rates, and water quality data of WWTPs at baseline period (2005-2017) are used for simulation of future period (2025-2099); (3) monthly simulations are performed for both baseline and future periods (Near Future (2025–2049), Mid Future (2050–2074), and Far Future (2075–2099)). Since the model was already shown to perform adequately, the parameter ranges calibrated in Vagheei et al. (2022) are also used for the present work. The SUFI-2 algorithm provided in the SWAT-CUP (SWAT Calibration and Uncertainty Program) tool is used to generate 95PPUs (the 95 % prediction uncertainty) of discharge and concentrations of nitrate, ammonium,

and total phosphorus for the baseline and future scenarios. Nitrate concentrations generated with the model are then coupled to the regression equation obtained in Vagheei et al. (2022) to produce the time series of IBMWP values. As described in Vagheei et al. (2022), this equation is resulted from the linear regression analysis between concentrations of nitrate and values of IBMWP resulting from sampling campaigns. Finally, the ecological status of rivers is determined based on projected chemical and biological status (using nitrate, ammonium, and total phosphorus as well as the macroinvertebrate community (the IBMWP index)) according to the Spanish Royal Decree 817/2015, that follows the "OOAO" principle. This principle is based upon the assumption that the worst status of the quality indicators used in the assessment determines the final ecological status (Zacharias et al., 2020). A complete description of the ecological status assessment can be found in Vagheei et al. (2022).

2.5. Quantification of uncertainties

Uncertainties of discharge, nitrate, ammonium, total phosphorus, and IBMWP projections are analyzed for the whole future period (2025–2099) to compare the contribution of three types of uncertainty, i.e., hydrologic parameter uncertainty (the differences between projections resulting from 500 SUFI-2 parameter solutions sets), GCM uncertainty (the differences between projections based on the use of five GCMs), and emission uncertainty (the differences between projections resulting from the use of four RCPs). For this purpose, projections are first averaged over time (900 months), and uncertainty analysis is then performed on these temporally averaged values as described below:

- 1. Hydrologic parameter uncertainty is quantified by averaging projected values over four RCPs and five GCMs and calculating the standard deviation over 500 (SUFI-2 parameter solutions sets) values.
- GCM uncertainty is quantified by averaging projected values over four RCPs and 500 SUFI-2 parameter solutions sets and calculating the standard deviation over five GCMs values.
- Emission uncertainty is quantified by averaging projected values over five GCMs and 500 SUFI-2 parameter solutions sets and calculating the standard deviation over four RCPs values.

3. Results and discussion

3.1. Projected temperature, precipitation, and discharge

Albaida Valley is expected to face rising temperature and reduced precipitation and discharge in all possible emission scenarios (Tables S2 and S3 in the Supplement). Detailed explanations of the projected temperature, precipitation, and discharge are provided in the Supplementary material.

3.2. Projected chemical indicators

95PPUs of chemical indicators of 14 representative sites were generated for baseline and future periods. Fig. S8 in the Supplement, as an example, shows the generated 95PPUs of nitrate load at the Site P14 for the baseline and the future (for GCM1 under RCP2.6 scenario). Simulated concentrations of nitrate, ammonium, and total phosphorus within the Albaida Valley

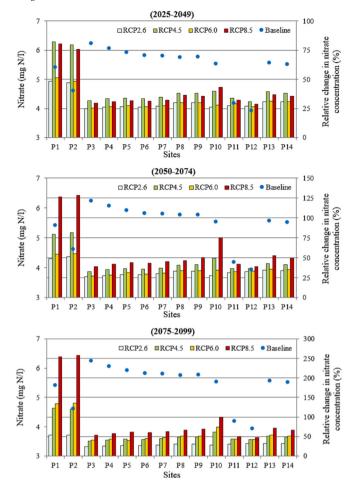


Fig. 3. Change in average nitrate concentration (%) at 14 representative sites compared to the baseline (2005–2017) using ensemble of five GCMs for the Near Future (2025–2049), Mid Future (2050–2074), and Far Future (2075–2099) for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenarios.

are projected to change in future compared to the baseline. According to the Figs. S9 to S13 in the Supplement, certain sites (i.e., Sites P1-P2, and P10-P12) have high variability of nitrate status (high uncertainty) in both baseline and future periods, where their nitrate status ranges from moderate to high. For the rest of sites (i.e., Sites P3-P9, and P13-P14), while nitrate status ranges from moderate to good in baseline period (mid variability), it is mostly moderate (low uncertainty) in the future. As Figs. S14 to S18 in the Supplement show, ammonium status of certain sites located in the upstream of rivers (i.e., Sites P1 and P2 in the Clariano River, and the Site P10 in the Albaida River) is always high (low variability) in both baseline and future periods. While for the rest of sites, ammonium status is mostly moderate with the probability of good and high status at some points, as well. Figs. S19 to S23 in the Supplement also represent that the total phosphorus status ranges from moderate and high at certain sites (i.e., Sites P1, P2, and P10). The average values are very representative for the rest of sites as uncertainty does not change too much and the total phosphorus status is moderate at these sites. Ensemble average nitrate concentration of 14 representative sites is projected to increase compared to the baseline period in all scenarios (Fig. 3). The most noticeable increases are found at P1, P2, and P10, where flow rates are low and hence less dilution is expected. Fig. 4 indicates that ensemble average ammonium concentration within the valley will likely increase compared to the baseline period in all sites. Ensemble average total phosphorus concentration within the valley is also predicted to increase compared to the baseline period along the streams of the valley apart from P1 and P2 (Fig. 5). These results are in agreement with predictions of changes in instream nutrient

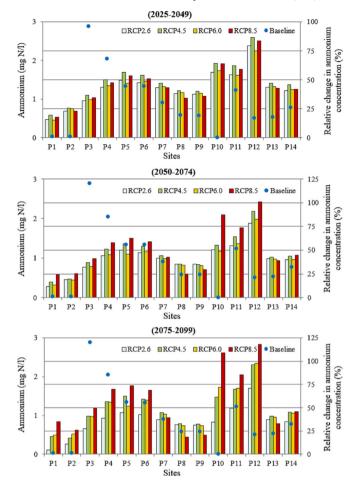


Fig. 4. Change in average ammonium concentration (%) at 14 representative sites compared to the baseline (2005–2017) using ensemble of five GCMs for the Near Future (2025–2049), Mid Future (2050–2074), and Far Future (2075–2099) for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenarios.

concentrations (Whitehead et al., 2006; Jeppesen et al., 2011), particularly with studies which suggest increases in riverine phosphorus and nitrogen concentrations (Arheimer et al., 2005; Kaste et al., 2006; Atkins, 2014; Molina-Navarro et al., 2014; Charlton et al., 2018). In fact, lower river flow under climate change reduces the dilution capacity of rivers, thereby resulting in higher concentrations of nutrients (Whitehead et al., 2006; Whitehead et al., 2009; Molina-Navarro et al., 2014; Charlton et al., 2018). Abily et al. (2021) discuss that climate change will lead to a dilution factor decrease for 11 % of the EU rivers. Baccour et al. (2021) also report the increased nitrate concentration and decreased water availability during droughts events in the Ebro River Basin (Spain), underlining the tradeoffs between quantity and quality of water. Using simulated concentrations and according to the Spanish Royal Decree 817/2015 and "OOAO" principle, moderate chemical status is predicted for all the representative sites in the future (Fig. 6), that emphasizes the evident need to control nutrient pollution.

3.3. Projected biological indicator

Values of IBMWP within the Albaida Valley are projected to change in future (Figs. S25 to S28 in the Supplement) compared to the baseline (Fig. S24 in the Supplement). As these figures indicate, the biological status, that ranges from bad to moderate at all the representative sites in the baseline period, remains bad to moderate at certain sites (i.e., sites P1-P2, and P10-P12) in the future, as well, and downgraded to the bad status (low uncertainty) at the rest of sites. Ensemble average IBMWP of 14 representative

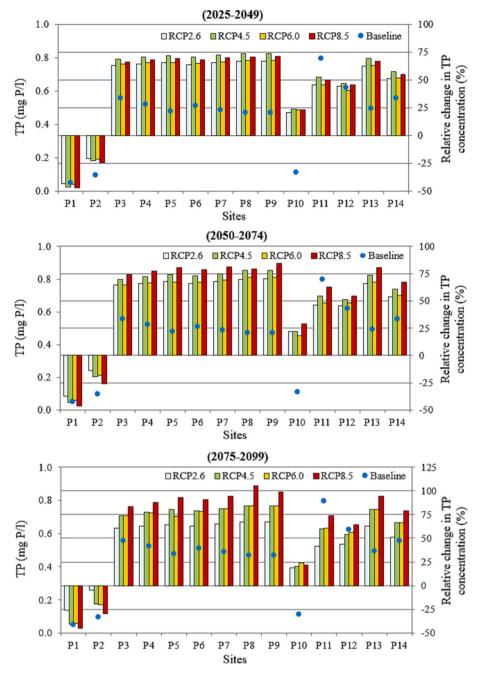


Fig. 5. Change in average total phosphorus concentration (%) at 14 representative sites compared to the baseline (2005–2017) using ensemble of five GCMs for the Near Future (2025–2049), Mid Future (2050–2074), and Far Future (2075–2099) for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenarios.

sites is predicted to decrease compared to the baseline period in all scenarios with highest decreases at sites that are already impacted (P3-P9; P13-P14) (Fig. 7). As expected, the highest reductions of IBMWP are projected for the most extreme scenario (i.e., RCP8.5). Using average values of IBMWP and according to the Spanish Royal Decree 817/2015, the biological status of certain representative sites may be downgraded from poor to bad status in the future (Fig. 8). The biological status of four representative sites (P3-P6) are classified as bad both in baseline and future periods. For the rest of representative sites (P1-P2; P7-P14) and under the RCP8.5 scenario, six sites (P7-P9; P12-P14) are expected to decrease their biological quality class in the Near and Mid Futures, while for the Far Future it is predicted that all 10 sites decrease their biological class and would be classified as bad. These results agree with other studies reporting that climate change may remarkably affect macroinvertebrates communities (Durance and Ormerod, 2007; Mantyka-Pringle et al., 2014; Kakouei et al., 2018). Therefore, the vulnerability of biodiversity to climate change should be reduced by conservation, protection, and restoration of freshwater ecosystems, as well as by targeted management to adapt to unavoidable effects of climate change (IPCC, 2022).

3.4. Quantified uncertainties

The quantified uncertainties for projected discharge (hydrologic parameter uncertainty: $0.07 m^3/s$, GCM uncertainty: $0.08 m^3/s$, and emission uncertainty: $0.04 m^3/s$), nitrate (hydrologic parameter uncertainty: 0.58 mg N/l, GCM uncertainty: 0.42 mg N/l, and emission uncertainty: 0.37 mg N/l), ammonium (hydrologic parameter uncertainty: 0.11 mg N/l), GCM uncertainty: 0.05 mg N/l, and emission uncertainty: 0.02 mg N/l), and emission uncertainty: 0.02 mg N/l), GCM uncertainty: 0.05 mg N/l, and emission uncertainty: 0.02 mg N/l), 0.02 mg N/l), and 0.02 mg N/l), $0.02 \text{$

Science of the Total Environment 893 (2023) 164645

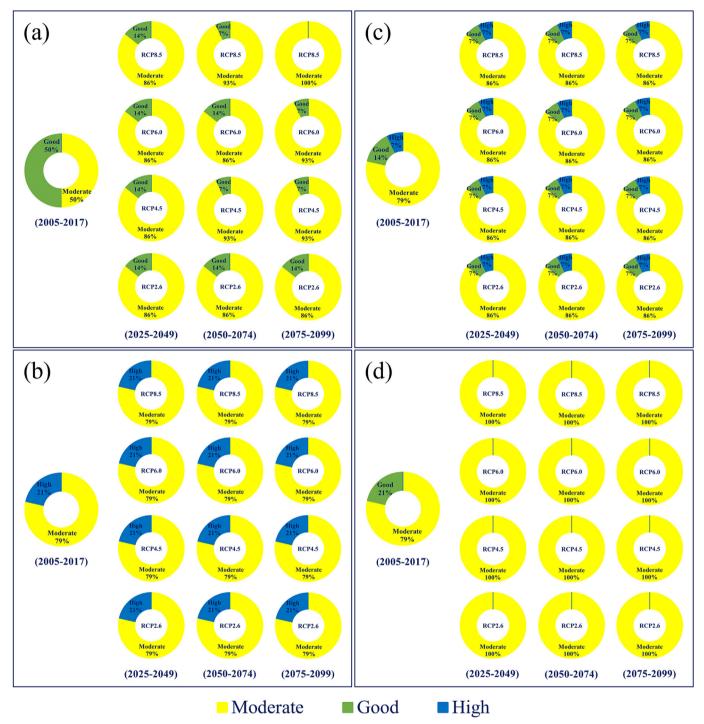


Fig. 6. Nitrate (a), ammonium (b), total phosphorus (c), and chemical (d) status of 14 representative sites in baseline and future periods for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenarios according to the Spanish Royal Decree 817/2015.

total phosphorus (hydrologic parameter uncertainty: 0.09 mg N/l, GCM uncertainty: 0.03 mg N/l, and emission uncertainty: 0.03 mg N/l), and IBMWP (hydrologic parameter uncertainty: 2.00, GCM uncertainty: 1.78, and emission uncertainty: 1.29), suggest that the hydrologic and GCM uncertainties are comparable for discharge and IBMWP. While for N and P, hydrologic uncertainty is the highest. Thus, the choice of the parameter values that describe the hydrologic properties of the catchment can affect the model predictions to an extent that is comparable or higher to the uncertainty of GCM predictions. This finding differs from that of Bennett et al. (2012) reporting that the most significant influence on uncertainty of

runoff projections in three watersheds in British Columbia is exerted by GCMs, followed by emissions scenarios and then hydrologic parameterizations. Similarly to the approach of Bennett et al. (2012), Table 2 reports the range of GCM uncertainties under the same RCPs. The table also reports the range of emission uncertainties for each GCM. The comparison of the values in Table 2 shows that GCM uncertainties are generally larger than emission uncertainties, that is in agreement with Bennett et al. (2012). Table S4 in the Supplement also presents the quantified hydrologic parameter uncertainties under each combination of GCM and RCP.

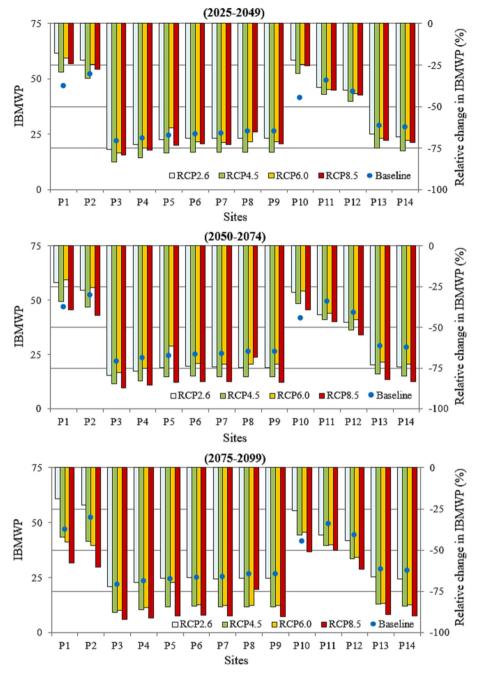


Fig. 7. Change in average IBMWP values (%) at 14 representative sites compared to the baseline (2005–2017) using ensemble of five GCMs for the Near Future (2025–2049), Mid Future (2050–2074), and Far Future (2075–2099) for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenarios.

3.5. Predicted ecological status

Based on predicted chemical and biological status and "OOAO" principle, it is calculated that the ecological status of most representative sites will worsen in the future compared to the baseline (Fig. 9, and Table S5 to Table S9 in the Supplement). For sites P1, P2, P10, and P11, the ecological status will remain poor under different scenarios except for the RCP8.5 scenario projecting bad status for all sites in the Far Future. The findings of this study are consistent with previous reports on the potential effects of climate change on freshwater ecosystems (Jeppesen et al., 2011; Mantyka-Pringle et al., 2014; Kakouei et al., 2018), particularly with the study predicting that the good ecological status of several EU rivers may be downgraded in future because of climate change, with more vulnerable sites located in Mediterranean countries (Abily et al., 2021). Interestingly, future nutrient loads in the Bellús Reservoir (located at downstream of the site P14) from the Albaida River are projected to be approximately similar to the baseline period (with a slight reduction tendency; Table S10 in the Supplement). However, a possible worsening of the ecological status of the reservoir is predicted for the future in case of a reduction in the volume of water stored in the reservoir due to increased evaporation and lower water availability, which would result in higher nutrient concentrations and increased eutrophication. This prediction agrees with Rocha et al. (2020), which states that climate change would negatively impact water quantity and quality in reservoirs. Other studies also reported higher concentrations of P and N in warm arid lakes during dry periods (when the water table is low) despite lower external nutrient loads due to enhanced evapotranspiration and decreased inflow (Özen et al., 2010; Jeppesen et al., 2011). Thus,

Science of the Total Environment 893 (2023) 164645

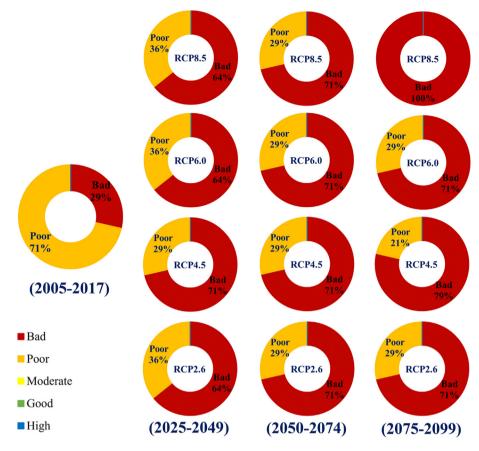


Fig. 8. Biological status of 14 representative sites in baseline and future periods for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenarios according to the Spanish Royal Decree 817/2015.

the improvement of WWTPs providing high-quality effluents and the sustainable management of agriculture with less nutrient loss to surface waters should be implemented in the valley to reduce the external nutrient load.

4. Conclusions

A SWAT-based ecohydrological modeling framework was used to investigate potential impacts of climate change on ecological status of Albaida Valley Rivers network using climate projections of five GCMs and four emission scenarios. Despite the different emission scenarios, Albaida Valley is expected to face rising temperature, reduced precipitation and discharge, and downgrading of ecological status. These findings emphasize the urgent need for scientifically informed decisions to manage and preserve freshwaters and address the importance of climate change mitigation and adaptation measures at the local, national, regional, and global levels. Even in case mitigation actions are implemented (e.g., for RCP2.6), this study projects that rivers may experience a lower ecological status even in the most optimistic scenario (RCP2.6). This suggests that adaptation measures (e.g., improving treatment of wastewater and/or adopting fertilizer management strategies to reduce nutrient leaching) will be necessary to avoid this situation. In addition, the uncertainty analysis performed in this

Table 2

GCM and emission uncertainties for projections of discharge (m^3/s) , nitrate (mg N/l), ammonium (mg N/l), total phosphorus (mg P/l), and IBMWP.

	GCM uncertainty						
	Under RCP2.6	Under RC	P4.5	Under RCP6.0	Under RCP8.5		
Discharge	0.11	0.06		0.09	0.07		
Nitrate	0.34	0.39		0.50	0.51		
Ammonium	0.05	0.05		0.06	0.06		
Total phosphorus	0.02	0.03		0.03	0.03		
IBMWP	2.18	1.27		2.38	1.71		
	Emission uncertainty						
	For GCM1	For GCM2	For GCM3	For GCM4	For GCM5		
Discharge	0.03	0.06	0.07	0.02	0.02		
Nitrate	0.44	0.31	0.39	0.43	0.37		
Ammonium	0.03	0.03	0.03	0.01	0.01		
Total phosphorus	0.03	0.04	0.04	0.03	0.03		
IBMWP	1.24	1.70	2.30	1.00	0.71		

Science of the Total Environment 893 (2023) 164645

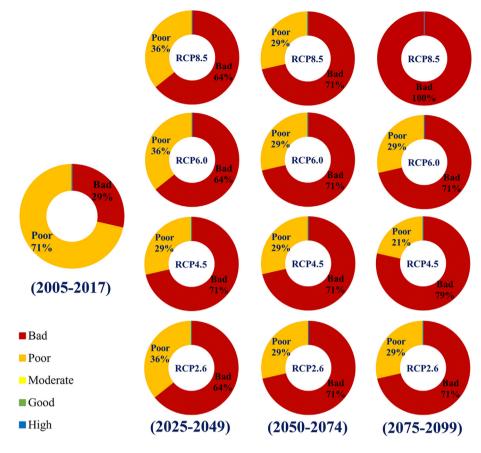


Fig. 9. Ecological status of 14 representative sites in baseline and future periods for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenarios using projected chemical and biological status and according to the Spanish Royal Decree 817/2015.

study reveals that the parameter values that characterize the catchment's hydrological features might have an equivalent level of impact on model predictions as GCMs, which this demonstrates the importance of characterizing the uncertainties from both hydrologic models and climate projections in climate change impact assessments. The modeling approach provided in this study could be generally used in different watersheds to investigate possible effects of changes in climate, land use and local management policies on freshwater ecosystems.

CRediT authorship contribution statement

Hamed Vagheei: Conceptualization, Methodology, Validation, Formal analysis, Investigation, Data curation, Writing – original draft, Writing – review & editing. Alex Laini: Conceptualization, Methodology, Writing – review & editing. Paolo Vezza: Conceptualization, Methodology, Writing – review & editing, Funding acquisition. Guillermo Palau-Salvador: Conceptualization, Methodology, Writing – review & editing. Fulvio Boano: Conceptualization, Methodology, Writing – review & editing, Supervision, Funding acquisition.

Data availability

Data will be made available on request.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Appendix A. Supporting information

Supplementary data to this article can be found online at https://doi.org/10.1016/j.scitotenv.2023.164645.

References

- Abbaspour, K.C., 2015. SWAT-CUP: SWAT Calibration and Uncertainty Programs A User Manual. Duebendorf, Switz. Eawag: Swiss Federal Institute of Aquatic Science and Technology, Duebendorf, Switzerland, p. 100. https://swat.tamu.edu/media/114860/ usermanual.swatcup.pdf.
- Abbaspour, K.C., Rouholahnejad, E., Ashraf Vaghefi, S., Srinivasan, R., Yang, H., Kløve, B., 2015. A continental-scale hydrology and water quality model for Europe: calibration and uncertainty of a high-resolution large-sale SWAT model. J. Hydrol. 524, 733–752. https://doi.org/10.1016/j.jhydrol.2015.03.027.
- Abily, M., Acuña, V., Gernjak, W., Rodríguez-Roda, I., Poch, M., Corominas, L., 2021. Climate change impact on EU rivers' dilution capacity and ecological status. Water Res. 199, 117166. https://doi.org/10.1016/j.watres.2021.117166.
- Aghsaei, H., Mobarghaee Dinan, N., Moridi, A., Asadolahi, Z., Delavar, M., Fohrer, N., Wagner, P.D., 2020. Effects of dynamic land use/land cover change on water resources and sediment yield in the Anzali wetland catchment, Gilan, Iran. Sci. Total Environ. 712, 136449. https://doi.org/10.1016/j.scitotenv.2019.136449.
- Ahmadalipour, A., Moradkhani, H., Castelletti, A., Magliocca, N., 2019. Future drought risk in Africa: integrating vulnerability, climate change, and population growth. Sci. Total Environ. 662, 672–686. https://doi.org/10.1016/j.scitotenv.2019.01.278.
- Andersen, H.E., Kronvang, B., Larsen, S.E., Hoffmann, C.C., Jensen, T.S., Rasmussen, E.K., 2006. Climate-change impacts on hydrology and nutrients in a Danish lowland river basin. Sci. Total Environ. 365 (1–3), 223–237. https://doi.org/10.1016/j.scitotenv. 2006.02.036.
- Arheimer, B., Andreasson, J., Fogelberg, S., Johnsson, H., Pers, C.B., Persson, K., 2005. Climate change impact on water quality: model results from southern Sweden. Ambio 34, 559–566.
- Ashraf Vaghefi, S., Abbaspour, N., Kamali, B., Abbaspour, K.C., 2017. A toolkit for climate change analysis and pattern recognition for extreme weather conditions – case study: California-Baja California Peninsula. Environ. Model. Softw. 96, 181–198. https://doi. org/10.1016/j.envsoft.2017.06.033.

Atkins, 2014. Water Resources and Hydrogeology Framework (21981). Climate Change and Population Growth Modelling. Report for the Environment Agency.

- Aylward, B., Bandyopadhyay, J., Belausteguigotia, J.C., Borkey, P., Cassar, A.Z., Meadors, L., Saade, L., Siebentritt, M., Stein, R., Tognetti, S., Tortajada, C., 2005. Freshwater ecosystem services. Ecosystems and Human Well-being: Policy Responses. 3, pp. 213–255.
- Baccour, S., Albiac, J., Kahil, T., Esteban, E., Crespo, D., Dinar, A., 2021. Hydroeconomic modeling for assessing water scarcity and agricultural pollution abatement policies in the Ebro River Basin, Spain. J. Clean. Prod. 327, 129459. https://doi.org/10.1016/j. jclepro.2021.129459.

Bennett, K.E., Werner, A.T., Schnorbus, M., 2012. Uncertainties in hydrologic and climate change impact analyses in headwater basins of British Columbia. J. Clim. 25, 5711–5730.

- Birk, S., Bonne, W., Borja, A., Brucet, S., Courrat, A., Poikane, S., Solimini, A., van de Bund, W., Zampoukas, N., Hering, D., 2012. Three hundred ways to assess Europe's surface waters: an almost complete overview of biological methods to implement the Water Framework Directive. Ecol. Indic. 18, 31–41. https://doi.org/10.1016/j.ecolind.2011.10.009.
- Bo, T., Doretto, A., Laini, A., Bona, F., Fenoglio, S., 2017. Biomonitoring with macroinvertebrate communities in Italy: what happened to our past and what is the future. J. Limnol. 76 (Suppl. 1), 21–28. https://doi.org/10.4081/jlimnol.2016.1584.
- Bouraoui, F., Galbiati, L., Bidoglio, G., 2002. Climate change impacts on nutrient loads in the Yorkshire Ouse catchment (UK). Hydrol. Earth Syst. Sci. 6 (2), 197–209. https://doi.org/ 10.5194/hess-6-197-2002.

Canter, L.W., 2018. River Water Quality Monitoring. CRC Press.

- Carr, M., Li, L., Sadeghian, A., Phillips, I.D., Lindenschmidt, K.-E., 2019. Modelling the possible impacts of climate change on the thermal regime and macroinvertebrate species of a regulated prairie river. Ecohydrology 12, e2102. https://doi.org/10. 1002/eco.2102.
- Charlton, M.B., Bowes, M.J., Hutchins, M.G., Orr, H.G., Soley, R., Davison, P., 2018. Mapping eutrophication risk from climate change: future phosphorus concentrations in English rivers. Sci. Total Environ. 613–614, 1510–1526. https://doi.org/10.1016/j.scitotenv.2017. 07.218.
- Deser, C., Phillips, A., Bourdette, V., Teng, H., 2012. Uncertainty in climate change projections: the role of internal variability. Clim. Dyn. 38, 527–546. https://doi.org/10.1007/ s00382-010-0977-x.
- Durance, I., Ormerod, S.J., 2007. Climate change effects on upland stream macroinvertebrates over a 25-year period. Glob. Chang. Biol. 13, 942–957. https://doi.org/10.1111/j.1365-2486.2007.01340.x.
- Fader, M., Shi, S., von Bloh, W., Bondeau, A., Cramer, W., 2016. Mediterranean irrigation under climate change: more efficient irrigation needed to compensate for increases in irrigation water requirements. Hydrol. Earth Syst. Sci. 20, 953–973. https://doi.org/10. 5194/hess-20-953-2016.
- Gaitán, E., Monjo, R., Pórtoles, J., Pino-Otín, M.R., 2020. Impact of climate change on drought in Aragon (NE Spain). Sci. Total Environ. 740, 140094. https://doi.org/10.1016/j. scitotenv.2020.140094.
- Goodarzi, M.R., Vagheei, H., Mohtar, R.H., 2020. The impact of climate change on water and energy security. Water Supply 20 (7), 2530–2546. https://doi.org/10.2166/ws.2020. 150.
- Guareschi, S., Laini, A., Sanchez-Montoya, M.M., 2017. How do low-abundance taxa affect river biomonitoring? Exploring the response of different macroinvertebrate-based indices. J. Limnol. 76 (Suppl. 1), 9–20.
- Guse, B., Kail, J., Radinger, J., Schröder, M., Kiesel, J., Hering, D., Wolter, C., Fohrer, N., 2015. Eco-hydrologic model cascades: simulating land use and climate change impacts on hydrology, hydraulics and habitats for fish and macroinvertebrates. Sci. Total Environ. 533, 542–556. https://doi.org/10.1016/j.scitotenv.2015.05.078.
- Hargreaves, G.L., Hargreaves, G.H., Riley, J.P., 1985. Agricultural benefits for Senegal River Basin. J. Irrig. Drain. Eng. 111 (2), 113–124.
- Hempel, S., Frieler, K., Warszawski, L., Schewe, J., Piontek, F., 2013. A trend-preserving bias correction - the ISI-MIP approach. Earth Syst. Dyn. 4 (2), 219e236.
- Holguin-Gonzalez, J.E., Boets, P., Alvarado, A., Cisneros, F., Carrasco, M.C., Wyseure, G., Nopens, I., Goethals, P.L.M., 2013. Integrating hydraulic, physicochemical and ecological models to assess the effectiveness of water quality management strategies for the River Cuenca in Ecuador. Ecol. Model. 254, 1–14. https://doi.org/10.1016/j.ecolmodel.2013. 01.011.
- IPCC, 2013. In: Stocker, T.F., Qin, D., Plattner, G.-K., Tignor, M., Allen, S.K., Boschung, J., Nauels, A., Xia, Y., Bex, V., Midgley, P.M. (Eds.), Climate Change 2013: The Physical Science Basis. Contribution of Working Group 1 to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change. Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA 1535 pp. https://www.ipcc.ch/site/assets/uploads/ 2018/02/WG1AR5_all_final.pdf.
- IPCC, 2022. In: Pörtner, H.-O., Roberts, D.C., Poloczanska, E.S., Mintenbeck, K., Tignor, M., Alegría, A., Craig, M., Langsdorf, S., Löschke, S., Möller, V., Okem, A. (Eds.), Summary for Policymakers. Climate Change 2022: Impacts, Adaptation, and Vulnerability. Contribution of Working Group II to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change [H.-O. Pörtner, D.C. Roberts, M. Tignor, E.S. Poloczanska, K. Mintenbeck, A. Alegría, M. Craig, S. Langsdorf, S. Löschke, V. Möller, A. Okem, B. Rama (eds.)]. Cambridge University Press In Press.
- Jeppesen, E., Kronvang, B., Olesen, J.E., et al., 2011. Climate change effects on nitrogen loading from cultivated catchments in Europe: implications for nitrogen retention, ecological state of lakes and adaptation. Hydrobiologia 663, 1–21. https://doi.org/10.1007/ s10750-010-0547-6.
- Jun, X., Shubo, C., Xiuping, H., Rui, X., Xiaojie, L., 2010. Potential impacts and challenges of climate change on water quality and ecosystem: case studies in representative rivers in China. J. Resour. Ecol. 1 (1), 31–35. https://doi.org/10.3969/j.issn.1674-764x.2010. 01.004.
- Kakouei, K., Kiesel, J., Domisch, S., Irving, K.S., Jähnig, S.C., Kail, J., 2018. Projected effects of climate change-induced flow alterations on stream macroinvertebrate abundances. Ecol. Evol. 8, 3393–3409. https://doi.org/10.1002/ece3.3907.

- Kaste, O., Wright, R.F., Barkved, L.J., Bjerkeng, B., Engen- Skaugen, T., Magnusson, J., Saelthun, N.R., 2006. Linked models to assess the impacts of climate change on nitrogen in a Norwegian river basin and fjord system. Sci. Total Environ. 365, 200–222.
- Kundzewicz, Z.W., Krysanova, V., Benestad, R.E., Hov, Ø., Piniewski, M., Otto, I.M., 2018. Uncertainty in climate change impacts on water resources. Environ. Sci. Pol. 79, 1–8. https://doi.org/10.1016/j.envsci.2017.10.008.
- Lei, C., Wagner, P.D., Fohrer, N., 2022. Influences of land use changes on the dynamics of water quantity and quality in the German lowland catchment of the Stör. Hydrol. Earth Syst. Sci. 26, 2561–2582. https://doi.org/10.5194/hess-26-2561-2022.
- López-López, E., Sedeño-Díaz, J.E., Mendoza-Martínez, E., Gómez-Ruiz, A., Martínez Ramírez, E., 2019. Water quality and macroinvertebrate community in dryland streams: the case of the Tehuacán-Cuicatlán Biosphere Reserve (México) facing climate change. Water 11, 1376. https://doi.org/10.3390/w11071376.
- Lutz, A.F., ter Maat, H.W., Wijngaard, R.R., et al., 2019. South Asian river basins in a 1.5 °C warmer world. Reg. Environ. Chang. 19, 833–847. https://doi.org/10.1007/s10113-018-1433-4.
- Mahmoodi, N., Wagner, P.D., Kiesel, J., Fohrer, N., 2021. Modeling the impact of climate change on streamflow and major hydrological components of an Iranian Wadi system. J. Water Clim. Chang. 12 (5), 1598–1613. https://doi.org/10.2166/wcc. 2020.098.
- Mantyka-Pringle, C.S., Martin, T.G., Moffatt, D.B., Linke, S., Rhodes, J.R., 2014. Understanding and predicting the combined effects of climate change and land-use change on freshwater macroinvertebrates and fish. J. Appl. Ecol. 51, 572–581. https://doi.org/10.1111/ 1365-2664.12236.
- Marcinkowski, P., Piniewski, M., Kardel, I., Szcześniak, M., Benestad, R., Srinivasan, R., Ignar, S., Okruszko, T., 2017. Effect of climate change on hydrology, sediment and nutrient losses in two lowland catchments in Poland. Water 9, 156. https://doi.org/10.3390/ w9030156.
- Molina-Navarro, E., Trolle, D., Martínez-Pérez, S., Sastre-Merlín, A., Jeppesen, E., 2014. Hydrological and water quality impact assessment of a Mediterranean limno-reservoir under climate change and land use management scenarios. J. Hydrol. 509, 354–366. https://doi.org/10.1016/j.jhydrol.2013.11.053.
- Morales-Marín, L.A., Rokaya, P., Sanyal, P.R., Sereda, J., Lindenschmidt, K.E., 2019. Changes in streamflow and water temperature affect fish habitat in the Athabasca River basin in the context of climate change. Ecol. Model. 407, 108718. https://doi.org/10.1016/j. ecolmodel.2019.108718.
- Moustakis, Y., Papalexiou, S.M., Onof, C.J., Paschalis, A., 2021. Seasonality, intensity, and duration of rainfall extremes change in a warmer climate. Earth's Future 9, e2020EF001824. https://doi.org/10.1029/2020EF001824.
- Özen, A., Karapınar, B., Kucuk, İ., Jeppesen, E., Beklioglu, M., 2010. Drought-induced changes in nutrient concentrations and retention in two shallow Mediterranean lakes subjected to different degrees of management. Hydrobiologia 646, 61–72. https://doi.org/10.1007/ s10750-010-0179-x.
- Park, J.Y., Park, G.A., Kim, S.J., 2013. Assessment of future climate change impact on water quality of Chungju Lake, South Korea, using WASP coupled with SWAT. J. Am. Water Resour. Assoc. 49 (6), 1225–1238. https://doi.org/10.1111/jawr.12085.
- Peña-Angulo, D., Vicente-Serrano, S.M., Domínguez-Castro, F., Noguera, I., Tomas-Burguera, M., López-Moreno, J.I., Lorenzo-Lacruz, J., El Kenawy, A., 2021. Unravelling the role of vegetation on the different trends between climatic and hydrologic drought in headwater catchments of Spain. Anthropocene 36, 100309. https://doi.org/10.1016/j.ancene.2021. 100309.
- Perez, J., Menendez, M., Mendez, F.J., Losada, I.J., 2014. Evaluating the performance of Cmip3 and Cmip5 global climate models over the North-East Atlantic Region. Clim. Dyn. 43, 2663–2680. https://doi.org/10.1007/s00382-014-2078-8.
- Povilaitis, A., Widén-Nilsson, E., Sarauskienė, D., Kriaučiūnienė, J., Jakimavičius, D., Bukantis, A., Kazys, J., Lozys, L., Kesminas, V., Virbickas, T., Pliūraitė, V., 2018. Potential impact of climate change on nutrient loads in Lithuanian rivers. Environ. Eng. Manag. J. 17 (9), 2229–2240.
- Qiu, J., Shen, Z., Xie, H., 2023. Drought impacts on hydrology and water quality under climate change. Sci. Total Environ. 858, 159854. https://doi.org/10.1016/j.scitotenv. 2022.159854.
- Rocha, J., Carvalho-Santos, C., Diogo, P., Beça, P., Keizer, J.J., Nunes, J.P., 2020. Impacts of climate change on reservoir water availability, quality and irrigation needs in a water scarce Mediterranean region (southern Portugal). Sci. Total Environ. 736, 139477. https://doi.org/10.1016/j.scitotenv.2020.139477.
- Sellami, H., Benabdallah, S., La Jeunesse, I., Vanclooster, M., 2016. Climate models and hydrological parameter uncertainties in climate change impacts on monthly runoff and daily flow duration curve of a Mediterranean catchment. Hydrol. Sci. J. 61, 1415–1429.
- Semenov, M.A., Stratonovitch, P., 2010. Use of multi-model ensembles from global climate models for assessment of climate change impacts. Clim. Res. 41, 1–14. https://doi.org/ 10.3354/cr00836.
- Shen, M., Chen, J., Zhuan, M., Chen, H., Xu, Ch., Xiong, L., 2018. Estimating uncertainty and its temporal variation related to global climate models in quantifying climate change impacts on hydrology. J. Hydrol. 556, 10–24. https://doi.org/10.1016/j.jhydrol.2017.11. 004.
- Shrestha, R.R., Dibike, Y.B., Prowse, T.D., 2012. Modeling climate change impacts on hydrology and nutrient loading in the Upper Assiniboine catchment. J. Am. Water Resour. Assoc. 48, 74–89. https://doi.org/10.1111/j.1752-1688.2011.00592.x.
- Shrestha, S., Bhatta, B., Shrestha, M., Shrestha, P.K., 2018. Integrated assessment of the climate and land use change impact on hydrology and water quality in the Songkhram River Basin, Thailand. Sci. Total Environ. 643, 1610–1622. https://doi.org/10.1016/j. scitotenv.2018.06.306.
- Spanish Royal Decree 817/2015. Royal Decree 817/2015 of 11 September, Establishing the Criteria for Monitoring and Evaluating the Status of Surface Waters and Environmental Quality Standards. BOE 219, 80582–80672. https://www.boe.es/eli/es/rd/2015/09/ 11/817.

H. Vagheei et al.

- Strobl, R.O., Robillard, P.D., 2008. Network design for water quality monitoring of surface freshwaters: a review. J. Environ. Manag. 87 (4), 639–648. https://doi.org/10.1016/j. jenvman.2007.03.001.
- Tabari, H., 2020. Climate change impact on flood and extreme precipitation increases with water availability. Sci. Rep. 10, 13768. https://doi.org/10.1038/s41598-020-70816-2.
- Tang, Y., Marshall, L., Sharma, A., Ajami, H., 2019. Modelling precipitation uncertainties in a multi-objective Bayesian ecohydrological setting. Adv. Water Resour. 123, 12–22. https://doi.org/10.1016/j.advwatres.2018.10.0.
- Tarekegn, N., Abate, B., Muluneh, A., Dile, Y., 2022. Modeling the impact of climate change on the hydrology of Andasa watershed. Model. Earth Syst. Environ. 8, 103–119. https://doi.org/10.1007/s40808-020-01063-7.
- Teutschbein, C., Seibert, J., 2012. Bias correction of regional climate model simulations for hydrological climate-change impact studies: review and evaluation of different methods. J. Hydrol. 456–457, 12–29. https://doi.org/10.1016/j.jhydrol.2012.05.052.
- Theodoropoulos, C., Karaouzas, I., 2021. Climate change and the future of Mediterranean freshwater macroinvertebrates: a model-based assessment. Hydrobiologia 848, 5033–5050. https://doi.org/10.1007/s10750-021-04691-x.
- Tueros, I., Borja, Á., Larreta, J., Rodríguez, J.G., Valencia, V., Millán, E., 2009. Integrating long-term water and sediment pollution data, in assessing chemical status within the European Water Framework Directive. Mar. Pollut. Bull. 58 (9), 1389–1400. https:// doi.org/10.1016/j.marpolbul.2009.04.014.
- Tzanakakis, V.A., Paranychianakis, N.V., Angelakis, A.N., 2020. Water supply and water scarcity. Water 12, 2347. https://doi.org/10.3390/w12092347.

- UN-Water, 2020. United Nations World Water Development Report 2020: Water and Climate Change. UNESCO, Paris.
- Vagheei, H., Laini, A., Vezza, P., Palau-Salvador, G., Boano, F., 2022. Ecohydrologic modeling using nitrate, ammonium, phosphorus, and macroinvertebrates as aquatic ecosystem health indicators of Albaida Valley (Spain). J. Hydrol. Reg. Stud. 42, 101155. https:// doi.org/10.1016/j.ejrh.2022.101155.
- Vitecek, S., Johnson, R.K., Poikane, S., 2021. Assessing the ecological status of European rivers and lakes using benthic invertebrate communities: a practical catalogue of metrics and methods. Water 13, 346. https://doi.org/10.3390/w13030346.
- Wang, K., Davies, E.G.R., Liu, J., 2019. Integrated water resources management and modeling: a case study of Bow River basin, Canada. J. Clean. Prod. 240, 118242. https://doi. org/10.1016/j.jclepro.2019.118242.
- Whitehead, P.G., Wilby, R.L., Butterfield, D., Wade, A.J., 2006. Impacts of climate change on in-stream nitrogen in a lowland chalk stream: an appraisal of adaptation strategies. Sci. Total Environ. 365 (1–3), 260–273. https://doi.org/10.1016/j.scitotenv.2006.02.040.
- Whitehead, P.G., Wilby, R.L., Battarbee, R.W., Kernan, M., Wade, A.J., 2009. A review of the potential impacts of climate change on surface water quality. Hydrol. Sci. J. 54 (1), 101–123. https://doi.org/10.1623/hysj.54.1.101.
- Zacharias, I., Liakou, P., Biliani, I., 2020. A review of the status of surface European waters twenty years after WFD introduction. Environ. Process. 7, 1023–1039. https://doi.org/ 10.1007/s40710-020-00458-z.