

Climate change effects on the hydrological regime in non perennial river basins

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Abstract

Recent years have witnessed an increasing interest on global climate change and, although we are only at the first stage of the projected trends, some signals of climate alteration are already visible. Climate change encompasses modifications in the characteristics of several interrelated climate variables, and unavoidably produces relevant effects on almost all the natural processes related to the hydrological cycle. This study focuses on the analysis of potential impacts of climate variations on the streamflow regime of small river basins in Mediterranean seasonally dry regions. Such areas, in fact, could be particularly vulnerable to climate change due to the general water scarcity and intermittence of streamflow, which are strongly controlled by the inter-annual and seasonal variability of precipitation and evapotranspiration.

This paper attempts to provide a quantitative evaluation of potential variations in the flow duration curves (FDCs) and in the partitioning between surface and subsurface contributions to streamflow, induced by climate changes projected over the next century in five different river basins, within the northwestern Sicily (Italy), representative of wide hydrological behaviors. To this aim, it is here used a recent hydrological model, the ModABa (*MODEL for Annual flow duration curves assessment in ephemeral small BASins*), which is first calibrated

30 at each site with regard to a past period with available streamflow observations, and, then, is
31 forced by daily precipitation and reference evapotranspiration series representative of the
32 current climatic conditions (2003-2013) and two future temporal horizons, referring to the
33 time windows 2045-2065 and 2081-2100 and centered around 2055 and 2090, respectively.
34 Current and future climatic series are generated by a weather generator, the AWE-GEN
35 (*Advanced WEather GENerator*), based on a stochastic downscaling of an ensemble of
36 General Circulation Models (GCMs).

37 The results quantify how the projected reduction in precipitation and increase in
38 temperature are reflected in the hydrological response of the selected basins through a
39 sensible downshift of the FDCs and a significant reduction in the mean annual streamflow.
40 Substantial alterations in streamflow seasonality and in the relative importance of the surface
41 and subsurface components are also highlighted. Results indicate how the torrential character
42 of Sicilian river basins could be exacerbated by climate change, with a lengthening of the dry
43 season and a higher intensity of extreme floods due to intensive rainfall events. All this could
44 have important implications for the water resource management and for the sustainability of
45 many riparian Mediterranean ecosystems.

46 **Keywords:** Climate Change; FDC; ModABa; GCM; Ephemeral basins

47 **1 Introduction**

48 Knowledge about flow regime in river basins has a capital importance for a variety of
49 practical applications, especially for watershed management and water sustainable use. The
50 hydrological regime plays a crucial role in determining the biotic composition, structure and
51 function of river basins. A full understanding of surface water availability and seasonality
52 should drive us towards a rational use of the water resource, aimed to satisfy anthropogenic
53 needs, warranting, at the same time, sufficient resources to support ecosystems services.

54 Streamflow regime in a river is highly dependent on different climatic factors, among
55 which the most important is surely the precipitation, in terms of frequency, intensity and
56 seasonal distribution of rainfall events. The cause-effect relationship between precipitation
57 and discharge becomes more noticeable in non perennial rivers, where streamflow mainly
58 relies on surface runoff. Thus, the study of the hydrological regime under transient climate
59 conditions results particularly relevant in ephemeral river basins, such as those characterizing
60 many arid and semi-arid regions, where the variations in streamflow characteristics strongly
61 depend on the underlying precipitation patterns.

62 Climate change is projected to have significant fallout upon streamflow regimes
63 worldwide. Since the beginning of climate change literature, researchers investigated on
64 potential impacts on streamflow regimes, which in turns could affect water resources
65 availability and riparian ecosystems existence in some vulnerable areas. Great attention has
66 been recently paid on this theme as demonstrated by the huge number of investigations
67 performed over the last decade (Arnell, 1999 and 2003; Middelkoop et al., 2001; Beyene et
68 al., 2010; Stahl et al., 2010; Bocchiola et al., 2011; Morán-Tejeda et al., 2011; Hannaford and
69 Buys, 2012, Confortola et al., 2013, Liuzzo et al., 2015), but there are few studies on this
70 specific topic with regard to ephemeral rivers (e.g., Ying and Huang, 1996; Schulze, 1997;
71 Batisani, 2011), where water scarcity frequently occur and where the effects of potential
72 changes in the climate forcings could imply eco-hydrological alterations heavier than for
73 perennial streams.

74 Future climatic projections, according to the *fifth assessment report* of the
75 *Intergovernmental Panel on Climate Change – IPCC-5AR* (IPCC, 2013), depict a changing
76 global water cycle, with increases in disparity between dry and wet regions. The
77 Mediterranean region, where future projections seem to lead toward an increase in aridity, can
78 be considered as a “hot spot” for climate change, with marked modifications of precipitation
79 in terms of both total amount and seasonal variability (e.g. Ulbrich et al., 2006; Sheffield and
80 Wood, 2008). In particular, in many warm Mediterranean areas, which are already
81 characterized by a limited water resource, the continue rise in greenhouse gases emission is
82 likely to induce a further reduction in water availability, also in the order of 20% (Mariotti et
83 al., 2008) or more, mainly due to the combined effect of a reduction in precipitation and a
84 simultaneous increment of temperature, with a consequent increment in the evaporation and
85 transpiration rates.

86 Starting from this premise, the primary objective of the present study is to explore the
87 climate change impact on the hydrological regime of ephemeral Mediterranean river basins. A
88 recent parsimonious hydrological model for the probabilistic characterization of the daily
89 streamflow in non perennial catchments, i.e. the ModABa - *MODEL for Annual flow duration*
90 *curves assessment in ephemeral small BASins* (Pumo et al., 2014) is used. The model
91 provides, for a given basin and climate forcings, the corresponding flow duration curve
92 (FDC), also giving useful indications about the repartition between the slow, subsurface, and
93 the fast, surface, components of streamflow. The FDC, which displays streamflow values
94 against their relative exceedance time, probably represents the most efficient method for

95 summarizing the hydrological features of a river, since it describes its ability in providing
96 flows of various magnitudes. A brief overview of the use and derivation of FDCs in
97 hydrology can be found in Vogel and Fennessey (1994, 1995). Here, an opportunely designed
98 module of the ModABa, i.e. *Module Statistics*, is used to infer the principal characteristics of
99 the hydrological regime from the FDCs.

100 To date, the ModABa has never been applied in basins different from that used for its
101 preliminary design and validation (i.e., Eleuterio at Lupo river basin, Sicily, Italy), where,
102 nevertheless, it has shown a relevant accuracy in reproducing the empirical FDC (Pumo et al.,
103 2014). The present paper then also represents a first attempt to apply the model to other
104 different basins, further testing its reliability and generality. In the previous application on the
105 Eleuterio at Lupo river basin, the reliability of the model was also successfully tested with
106 regard to different time periods characterized by a certain climatic variability, and this aspect
107 encouraged us to adopt the ModABa in the present work with the aim to investigate potential
108 modifications of the FDC under transient climate conditions.

109 Five river basins are here selected within the study area (i.e., northwestern Sicily, Italy); at
110 each site, the model is first calibrated against past streamflow observations, and then it is
111 forced by reliable stochastic current and future climatic scenarios in order to assess the impact
112 of climate change on the hydrological response.

113 The future climate scenarios are generated starting from precipitation and temperature
114 realizations obtained by General Circulation Models (GCMs). The GCM realizations here
115 considered correspond to two of the four greenhouse gas concentration trajectories adopted by
116 the IPCC-5AR (Moss et al., 2010; IPCC 2013): the *Representative Concentration Pathway*
117 (RCP) 4.5 and 8.5. The GCMs typically have a temporal and spatial resolution (daily or even
118 monthly time resolution and pixel dimension that ranges from about 100 to 300 km)
119 inadequate to represent hydrological processes at scales smaller than the continental scale, so
120 downscaling of GCM-based data is a key aspect in climatologically driven hydrological
121 simulation (Chen et al., 2011). The adoption of higher spatial resolution (e.g., 25–50 km)
122 regional climate models (RCMs) could still not provide the resolution required for robust
123 hydrological modeling as demonstrated by difference recent studies (e.g., Milly et al., 2005).
124 However, a wide range of statistical approaches (Wilby and Wigley, 1997), some of which
125 explicitly focused on hydrological applications (e.g., Fowler et al. 2007), can be applied to
126 downscale further to the local scale by relating point climate observations and climate models
127 outputs. In this study an opportune stochastic downscaling technique is adopted (Tebaldi et

128 al., 2004; 2005) using the Advanced WEather GENerator - AWE-GEN (Ivanov et al., 2007;
129 Fatichi et al., 2011). The AWE-GEN is a numerical tool capable of reproducing many
130 statistical properties of several meteorological variables from the hourly time-scale up to
131 several years. For each considered river basin, the *Baseline* scenario, representative of the
132 current climate (i.e., 2003-2013), is created, as well as two future climate scenarios for each
133 emission scenarios investigated (i.e., RCP 4.5 and 8.5), namely the climate projections at
134 2055 and 2090, based on an ensemble of GCMs projections over the periods 2045-2065 and
135 2081-2100, respectively. In particular, rainfall, temperature, solar radiation, relative humidity
136 and wind speed are simulated by the AWE-GEN in order to obtain stationary 50 years long
137 time series. Using the climatic series generated by the weather generator and taking also into
138 account the possible effect of CO₂ increase on stomatal resistance, the FAO-56 Penman-
139 Monteith method (Allen et al. 1998) is used to estimate reference evapotranspiration series for
140 each scenario.

141 The paper is organized as follows. In Section 2 the study area, with the five selected river
142 basins, and the considered databases for the hydro-climatic data are described. It is also
143 provided a description of both the adopted hydrological model, including model calibration at
144 each site, and the weather generator, including the used downscaling procedure. Section 3
145 presents the main results. More specifically, after a preliminary evaluation of the hydrological
146 model performances for the five basins in calibration, the principal differences emerging from
147 the comparison between the current and the future FDCs are analyzed, highlighting the
148 projected effects of climate change on the hydrological regime at each examined site. The
149 results achieved for the five cases here selected could be thought as representative of the
150 majority of Sicilian basins and, thus, they may be useful in the evaluation of the major
151 consequences of climate change at the regional level, with important implications that could
152 interest a wide range of sectors, from hydropower generation to agriculture, from industry to
153 domestic water use. This study may also provide a template for investigation on future water
154 resources in non perennial river basins under a changing climate, and, finally, could be
155 coupled with ecological models in order to furnish critical information about the conservation,
156 or the potential evolution, of the existing ecosystems.

157 **2 Data and methods**

158 **2.1 Study area**

159 The study area (Figure 1) is located in the northwestern Sicily (Italy). The precipitation
160 regime, according to the Köppen-Geiger climate classification, belongs to the warm temperate
161 class, with summer usually hot and dry, and with a maximum of precipitation in winter. The
162 climate and the morphology in this area concur in determining a prevalence of small river
163 basins characterized by non perennial hydrological regimes. Among these, five natural basins
164 with streams no subjected to regulation and representative of the entire study area, have been
165 selected mainly on the basis of the availability of long and reliable meteorological and
166 streamflow historical series; the sites also warrant the fulfilment of all the prerequisites for the
167 hydrological model adopted in this study (e.g., sub-daily time of concentration).

168 The selected river basins, whose elevation distribution and main river channels are
169 depicted in Figure 1, are: *Forgia at Lentina*, *Baiata at Sapone*, *Fastaia at La Chineia*, *Nocella*
170 *at Zucco* and *Senore at Finocchiara*. The main characteristics of each basin are synthesized in
171 Table 1, where it is reported the total area (ranging from 29 to 76 km²), the main river channel
172 length (from 7 to 24 km), the mean elevation (from 111 to 527 m a.s.l.) and the percentage of
173 impervious area (from 4.3 to 8.4%).

174 The topographic features have been derived by a DEM (*Digital Elevation Model*) with 100
175 m spatial resolution, while for the characterization of the soils and vegetations of each basin
176 and for the determination of the impervious portion, different sources have been selected and
177 crossed: the *Corinne Land Cover Map of Sicily* of 2006 (De Jacher, 2012), the *Soil Map of*
178 *Italy and Soil Data Base* of Italy's Soil Information System (SISI, www.soilmaps.it,
179 Costantini et al., 2012 and 2013) and the *Land Use Map of Sicily* at 1:250,000 scale (Regione
180 Siciliana, 1994).

181 The impervious part of the basins includes all the areas classified as urban, industrial and
182 infrastructural areas, roads and rural buildings, as well as very low permeable soils, with
183 outcrops and compact rocks. The spatial vegetation patterns for each basin (Figure 2a) have
184 been mainly derived from the *Corinne Land Cover Map*, distinguishing between three
185 different vegetational macro-classes, whose parameters are typical literature values used for
186 ecohydrological applications (e.g. Cuenca et al., 1997; Caylor et al., 2005; Pumo et al., 2014).
187 The adopted classification is essentially based on the rooting depth of vegetation; the three
188 classes correspond to areas mainly covered by deep roots plants, referred as trees (such as
189 broad-leaved, coniferous and mixed forests, woods and partially wooded land), medium deep

190 roots plants, referred as shrubs (e.g., shrubland and arable lands with small size tree crops
191 such as citrus groves, fruit trees or vineyard), and, finally, shallow roots plants, referred as
192 grasses (e.g., arable lands with herbaceous crop associations, grassland and sparsely vegetated
193 areas). The spatial soil patterns (Figure 2b), mainly derived through the *Soil Map of Italy*,
194 show the presence of seven different soil textural classes (according to USDA classification)
195 within the considered basins: sandy clay loam, sandy loam, silty clay, loam, silty loam, silty
196 clay loam and clay loam. Soil parameters for each textural class have been estimated using
197 data from the open source database of the National Centre for Soil Mapping - CNCP
198 (*Research Centre for Agrobiological and Pedology – CRA-ABP*, www.soilmaps.it) through
199 opportune pedotransfer functions suitable for Sicilian soils (Antinoro et al., 2008).

200 The relative percent cover of each vegetation and soil type present within the permeable
201 portion of the five basins and the corresponding characteristics, useful for the successive
202 application of the hydrological model here adopted, are listed in Tables 2a and 2b,
203 respectively; the reported values demonstrate how the choice of the case studies allows this
204 study to account for a good heterogeneity in terms of both vegetation coverage and soil type.

205 **2.2 Historical database**

206 Two different sources of data have been used, providing low (i.e. daily) and high (i.e.
207 hourly) frequencies climatic records which are necessary for the ModABa calibration and for
208 the weather generator set up, respectively.

209 The low frequency data have been provided by the OA-ARRA Regional Agency (*Agenzia*
210 *Regionale per i Rifiuti e le Acque*). The OA-ARRA archive provides daily streamflow series
211 and monthly areal precipitation for most of the Sicilian river basins, and, at the same time,
212 furnishes daily rainfall and temperature data collected by a dense network of gauging stations
213 over the territory. The five river basins selected for this study are indeed characterized by a
214 streamflow OA-ARRA gauge at the outlet and by the absence of human induced
215 modifications (e.g., dams, water diversions, etc.). Among the considered basins, the longest
216 available daily streamflow series is that relative to *Nocella at Zucco* (42 year, from 1958 to
217 2003, with 4 years gap) while the shortest series is that relative to *Senore at Finocchiara* (25
218 years from 1961 to 1986 with only 1 year gap). Because of the discontinuities, only a limited
219 part of the available dataset at each site has been considered, selecting the longest period with
220 continuous data as calibration period (Table 3). Thus, the size of the resulting calibration
221 periods ranges from 15 years (i.e., *Forgia at Lentina*) to 23 year (i.e., *Nocella at Zucco*) and,

222 for all the cases, it results sufficiently long to allow model calibration (Pumo et al., 2014). At
223 each site, areal estimates of daily rainfall and temperature relative to the calibration period
224 have been derived starting from the data collected by the OA-ARRA in 35 climatic stations
225 over the whole study area and using opportune spatial interpolation techniques. In particular,
226 algorithms analogous to those adopted by Di Piazza et al. (2011) and based on a distance
227 weighting scheme, which also take into account the elevation, have been here used at the
228 daily time scale. The reliability of such a procedure has been also successfully tested with
229 regard to rainfall by aggregating the daily areal estimates at the monthly scale and comparing
230 the resulting values with the monthly areal precipitation for each basin, separately retrieved
231 from the OA-ARRA dataset. Table 3 reports a list of all the raingauge (*P*) and thermometric
232 (*T*) stations considered in this study for the reconstruction of the historical series used in
233 calibration, while their locations are represented in Figure 1.

234 The weather generator here adopted needs different meteorological data at higher temporal
235 resolution (i.e., hourly) and relative to a recent and sufficient long period (e.g., about 10 years
236 or more) representative of the current climate conditions. The high frequency data have been
237 provided by the SIAS Regional Agency (*Servizio Informativo Agrometeorologico Siciliano*),
238 which furnishes hourly series of several climatic variables recorded since 2003 within the
239 Sicily. An internal (or the closest to the basin) station of the SIAS monitoring network has
240 been first selected for each river basin (Figure 1 and Table 3); the entire available hourly
241 series of precipitation, temperature, solar radiation, wind speed, air relative humidity and
242 pressure, recorded by each station until 2013, have been then retrieved and assumed as
243 representative of the entire basin.

244 **2.3 Hydrological model**

245 The adopted hydrological model is the ModABa - *MODEL for Annual flow duration curves*
246 *assessment in ephemeral BASins*, a model aimed to the reconstruction of the annual FDC of
247 the daily total streamflow (baseflow plus surface runoff) in catchments with sub-daily time of
248 concentration. One of the peculiarities of the model is that it can be also applied to cases of
249 non perennial hydrological regime, where streamflow seasonally becomes null for a
250 significant period of the year. The model, introduced in Pumo et al. (2014), has the great
251 advantage to be not computationally intensive, with few parameters that can be easily derived
252 from commonly available information about the soil and vegetation of a basin, and, moreover,
253 it can be applied simply using daily series of precipitation and temperature (or, in alternative,

254 reference evapotranspiration) as external forcings. The calibration procedure proposed by
255 Pumo et al. (2014) guarantees relevant model elasticity, meant as model ability in
256 reproducing, for the same basin, the empirical FDCs relative to different time periods and,
257 thus, characterized by a certain climatic variability. This aspect makes the model particularly
258 suitable for the purposes of this work. In what follows only the main features of the model are
259 recalled, while interested readers are referred to the original paper (Pumo et al., 2014) for a
260 more comprehensive discussion about some aspects here neglected or deeply synthesized.

261 The ModABa is a spatially lumped model and accounts for a different hydrological
262 behavior between the impervious (whose fraction is denoted as c_0) and the permeable (fraction
263 equal to $1-c_0$) portions of the basin. The total streamflow is assumed to be formed by two
264 contributions differently modelled: the slow subsurface contribution and the fast surface
265 contribution. The model conceptualizes the basin as a linear reservoir characterized by the
266 mean residence time of the basin with regard to the transferring process related to the slow
267 subsurface contribution, while, with regard to the surface contribution transferring process,
268 the basin is modeled as a linear channel which instantaneously transfers the generated fast
269 runoff towards the basin outlet.

270 The ModABa has a modular structure, which, in the original formulation (Pumo et al.,
271 2014), was essentially constituted by three interconnected modules:

- 272 1) *Module 1*: it identifies the *dry season*, i.e. part of the year with null streamflow, and the
273 *non-zero season*, i.e. remaining part of the year, and empirically provides the
274 probability of zero-flow within the considered basin as a function of the underlying
275 precipitation regime;
- 276 2) *Module 2*: it derives the non-zero FDC (FDC_{nz} , i.e. duration curve of the total
277 streamflow relative to the *non-zero season*), first separately estimating and then
278 overlapping the non-zero FDCs relative to the subsurface and the surface component of
279 streamflow ($SSFDC_{nz}$ and $SFDC_{nz}$, respectively);
- 280 3) *Module 3*: it combines the outputs of the previous two modules (i.e. the probability of
281 null streamflow and the FDC_{nz}) to obtain the annual FDC.

282 In view of the objectives of the present work, it has been here implemented and added a
283 new module, named *Module Statistics*, which computes: 1- the streamflow values (l/s) for a
284 set of characteristic percent durations; 2- the mean annual discharge, Q_m (l/s) and the mean
285 discharge over the only *non-zero season*, $Q_{m,nz}$ (l/s); 3- the mean durations in days of the *non-*

286 zero season (d_{nz}); 4- the mean annual runoff coefficient, (ϕ); 5- the mean annual runoff R
287 (mm/yr); 6- the surface component percent contribution to the mean annual runoff (SFC).

288 A schematic representation of the ModABa structure is provided in Figure 3 (from Pumo
289 et al., 2014). The first operational step consists in the separation of the hydrological year into
290 the *dry* and the *non-zero season*. This task is assigned to *Module 1* that, following an
291 approach analogous to that used in Croker et al. (2003), derives the probability that the
292 streamflow is null (p_{dry}) and, consequently, the mean duration of the *dry seasons* (d_{dry}). The
293 value of p_{dry} is estimated as a function of the mean annual values of precipitation (MAP) and
294 fraction of no rainy days (f_{NR}) by a multiple linear regressive model applied to log-
295 transformed data, according to the following power law:

$$296 \quad p_{dry} = m_0 \cdot MAP^{m_1} \cdot f_{NR}^{m_2} \quad (1)$$

297 where m_0 [year/mm], m_1 and m_2 are site-specific regression parameters. The *dry season* is the
298 driest period of the year of duration d_{dry} (Pumo et al., 2014); it is then assumed to be constant
299 in length, while its beginning and ending days can vary year by year.

300 Rainfall occurring over the *dry season* is assumed to form no significant streamflow and it
301 is therefore neglected by the model, while rainfall occurring during the remaining part of the
302 year can feed the two different contributions to total streamflow. The mechanisms of
303 streamflow generation are modelled by *Module 2*, which is, in turn, constituted by three sub-
304 modules, synthetically described below: *Sub-Module 2.1*, *2.2* and *2.3*.

305 *Sub-Module 2.1* is based on an ecohydrological model, originally introduced by Botter et
306 al. (2007 and 2008) and further extended by Pumo et al. (2013 and 2014), which is applied to
307 the permeable portion of the basin. Once the fraction of impervious area c_0 is derived, the
308 model requires the following spatially averaged parameters characterizing the mean soil and
309 vegetation of the permeable areas: the rooting depth, Z_r ; the threshold for canopy
310 interception, δ_{veg} ; the porosity, n ; the saturated hydraulic conductivity, k_s ; the relative soil
311 moisture contents at the wilting point (s_w) and at the field capacity (s_{fc}); the β coefficient used
312 to describe the hydraulic conductivity power law (Laio et al. 2001). The sub-module is based
313 on the analytical solution of a stochastic soil moisture balance equation. Water losses from the
314 soil, contributing to the baseflow, depend on the relative soil moisture content, whose
315 dynamics are, in turn, driven by external climatic forcings relative to the *non-zero season*.
316 Evapotranspiration is assumed to vanish below s_w and to increases linearly as the soil
317 moisture increases up to a maximum, potential, rate (ET_{max}) that is reached in correspondence

318 of s_{fc} . Field capacity is assumed as a threshold soil moisture content whose upcrossing
 319 determines the triggering mechanism for subsurface runoff generation. Rainfall interarrival
 320 process is simulated as a stationary marked Poisson process with rate λ_p (1/day) while the
 321 daily rainfall depths are assumed to be distributed according to an exponential distribution
 322 with mean $1/\gamma'_p$ (γ'_p in mm^{-1}). The process of rainfall interception by vegetation is also taken
 323 into account by a censoring process on the rainfall series that modifies the mean rainfall
 324 frequency in an apparent rate (λ'_p) on the basis of the canopy interception threshold, δ_{veg} .

325 The mean daily potential evapotranspiration, ET_{max} (mm/day) is derived from the reference
 326 evapotranspiration, ET_0 , as $ET_{max} = k_v \cdot ET_0$, where k_v is a constant referred as vegetational
 327 coefficient. Despite this procedure implies the introduction of a further parameter that has to
 328 be calibrated (i.e., k_v), it allows for the use of simple procedures, such as the Thornthwaite
 329 (1949) or the Hargreaves et al. (1985) methods, which offer the great advantage of being
 330 based only on the temperature information, which is often the only available climatic data.

331 The transport model (i.e., linear reservoir) adopted for *Sub-Module 2.1* assumes the
 332 residence times as exponentially distributed with parameter k , which is the inverse of the
 333 mean residence time of the basin. The SSFDC_{nz} is finally obtained as the complement of the
 334 cumulative exceeding probability of the subsurface streamflow, which is, in turn, derived
 335 from the steady state pdf of the specific subsurface streamflow, where this last is assumed as a
 336 Gamma-distribution (Pumo et al., 2014).

337 Relevant empirical relationships between the two calibration parameters of *Sub-Module*
 338 *2.1* (i.e., k and k_v) and the underlying climatic forcings were found in Pumo et al. (2013),
 339 where it was also demonstrated how the use of such relationships considerably improves
 340 model elasticity to climate variability. For this reason the model adopts two multiple linear
 341 regression models to assess such parameters as a function of λ'_p and ET_0 :

$$342 \begin{cases} k = a_1 + b_1 \cdot \lambda'_p + c_1 \cdot ET_0 \\ k_v = a_2 + b_2 \cdot \lambda'_p + c_2 \cdot ET_0 \end{cases} \quad (2)$$

343 where a_1 [1/day], b_1 , c_1 [1/mm], a_2 , b_2 [day], c_2 [day/mm] are site-specific regression
 344 parameters.

345 *Sub-Module 2.2* follows the same conceptualization of Yokoo and Sivapalan (2011). The
 346 surface component of streamflow is assumed to be formed by two different contributions: the
 347 *impervious runoff* and the *direct surface runoff*. The first is simply given by the entire amount
 348 of rain fallen onto the impervious portion of the basin during the *non-zero seasons*. The

349 second contribution derives from the permeable areas of the basin and accounts for a
 350 Hortonian infiltration-excess mechanism that could occur in presence of very intense rainfall
 351 events (heavy rains), when rainfall intensity is likely to exceed the soil infiltration capacity.
 352 The *direct surface runoff* is assumed to be formed by the excess of rainfall with respect to a
 353 critical heavy rain threshold, referred as *HRT* [mm/day]. *Sub-Module 2.2* computationally
 354 derives the $SFDC_{nz}$ directly from the non-zero precipitation duration curve (PDC_{nz}), through a
 355 not linear filter for the precipitation, whose parameters are c_0 and *HRT*. It is possible to obtain
 356 the surface contribution to the specific daily streamflow at the generic i -th day of the *non-zero*
 357 *seasons*, $Q_s(i)$, as a function of the precipitation at the same day, $R(i)$ (Pumo et al., 2014).
 358 Computing Q_s for each day of the *non-zero seasons*, the $SFDC_{nz}$ is finally derived as the
 359 complement of the cumulative “frequency of exceedance” distribution of Q_s .

360 In *Sub-Module 2.3*, the $SSFDC_{nz}$, resulting by *Sub-Module 2.1*, and the $SFDC_{nz}$, resulting
 361 by *Sub-Module 2.2*, are overlapped, deriving the non-zero total streamflow flow duration
 362 curve (FDC_{nz}). This last represents streamflow values occurring between 0 and p_{nz} , with $p_{nz} =$
 363 $1 - p_{dry}$. In order to obtain the annual FDC, it is necessary to rescale the interval $[0, p_{nz}]$ into
 364 the entire relative duration interval $[0, 1]$ by transforming the component flows of the FDC_{nz} .
 365 This task is assigned to *Module 3*, which is based on the theory of total probability. In
 366 particular, for the generic i -th streamflow Q_i , the transformed exceedance probability $P(Q_i)$
 367 over the interval $[0,1]$ is computed by the following equation:

$$368 \quad P(Q_i) = P(Q_i)_{nz} \cdot p_{nz} \quad (3)$$

369 where $P(Q_i)_{nz}$ is the exceedance probability of the streamflow Q_i derived by the FDC_{nz} .

370 **2.4 Model calibration**

371 The ModABa calibration at each basin has been performed, with regard to the calibration
 372 period previously selected (Sect. 2.2 and Table 3), adopting the same procedure used in Pumo
 373 et al. (2014). The first step is the determination of the impervious fraction of each basin and
 374 the estimation of the parameters representative of the mean vegetation and soil of the
 375 permeable portion of the basins. The information already discussed in Sect. 2.1 and
 376 synthesized by Tables 1 and 2 has been used to this aim, and, more specifically, the
 377 characteristics of the mean vegetation and soil have been obtained for each basin as weighted
 378 averages of the parameters relative to each type of vegetation and soil present within the
 379 basin, with weights given by the relative cover fractions. The estimated values for the five
 380 considered case studies are reported in Table 4.

381 The procedure for the assessment of the other calibration parameters can be summarized in
382 three sequential phases: 1. calibration of *Module 1* and, more specifically, of the regression
383 parameters of Eq. (1); 2. calibration of *Sub-Module 2.1* and, more specifically, of the
384 regression parameters of Eq. (2), considering a preliminary, empirically fixed, value of *HRT*;
385 3. optimization procedure for deriving the calibration parameter *HRT* of *Sub-Module 2.2*.

386 An opportune model performance index is required for the last two phases. To this aim it is
387 used the distance index Δ (Pumo et al., 2013; 2014), which is a measure of (dis)similarity
388 between corresponding empirical and theoretical FDCs whose value, ranging between 0 and
389 1, decreases as the matching between the two curves improves. The estimation of Δ follows a
390 simple distance-based procedure, with distances evaluated in fixed points (i.e., comparison
391 points) of the two compared curves.

392 The parameters of Eqs. (1) and (2) are estimated through a regression analysis that
393 considers different subperiods over the entire available dataset (i.e., calibration period). The
394 regression analysis for Eq. (1) explores the cause-effect relation between precipitation and
395 percentage of no-flow days. The empirical values of *MAP* and f_{NR} , and the corresponding
396 empirical mean annual percentage of days with null streamflow are first derived for all the
397 considered subperiods, and, then, interpolated by the power law of Eq. (1), assessing the
398 regression parameters m_0 , m_1 and m_2 by the least square method.

399 The regression analysis for Eq. (2) explores the relationships for each subperiod between
400 the underlying climatic parameters λ'_p and ET_0 , and two “optimal” values for k and k_v . Such
401 “optimal” parameters are values corresponding to the best model reproduction of the
402 empirical FDC at the level of each subperiod (i.e., minimum Δ). They are derived as best
403 solutions for each subperiod by a Monte Carlo procedure, after 200,000 simulations, each one
404 associated to a different combination of k and k_v randomly generated (i.e., randomly extracted
405 by a uniform distribution of each parameter within a wide range opportunely fixed). The
406 regression parameters of Eq. (2) between the “optimal” parameters k and k_v and the climatic
407 parameters λ'_p and ET_0 , are finally assessed, also in this case, by the least square method.

408 Regression analysis methods, such as those used for Eqs. (1) and (2), require to consider an
409 adequate number of sufficiently long subperiods within the entire calibration period; their
410 performances are, in fact, related to the explored variability of the independent variables (i.e.,
411 climatic forcings). The minimum size of the subperiods for Eqs. (1) and (2) calibration is
412 identified (Table 4) following the same approach adopted by Pumo et al. (2014), based on
413 Cochran’s formula (Cochran, 1977); all the continue subperiods within the calibration period

414 of each basin having size equal or longer than the minimum size found for Eq. (1) are
415 considered to calibrate *Module 1*, while for *Sub-Module 2.1*, the considered subperiods are
416 those generated by a moving time-slot over the calibration period with fixed size, coincident
417 with the minimum size found for Eq. (2).

418 The procedure adopted to calibrate Eq. (2) parameters requires to set a preliminary value
419 for the parameter *HRT* of *Sub-Module 2.2*, which also conditions the parameter γ'_p of *Sub-*
420 *Module 2.1*, representing an upper bound for infiltrating rainfall. The critical threshold *HRT*
421 [mm/day] defines the precipitation events that can be classified as “heavy rain”. In general,
422 the threshold value is related to the processes under analysis and the specific objectives of the
423 analysis and it depends on several basin characteristics (Pumo et al., 2014). It is often derived
424 from the cumulative frequency distribution of daily precipitation (PDC) as the daily
425 precipitation associated to a critical *i*-th percentile (Kunkel et al., 2007). Following the same
426 approach used in Pumo et al. (2014), the precipitation value corresponding to the 5-th
427 percentile of the PDC_{nz} (i.e., precipitation duration curve of the *non-zero season*) for each
428 investigated basin is here assumed as “first guess” value for *HRT* to be used in the Eq. (2)
429 calibration procedure. Nevertheless, the preliminary empirical estimation of *HRT* is
430 successively refined by an appropriate optimization procedure that represents the last
431 ModABa calibration phase. This is performed iteratively applying the model to the entire
432 calibration period with progressively increasing values of *HRT* and computing as final value
433 for *HRT* that associated to the minimum value of Δ (best model performances). In particular,
434 for each river basin, it has been considered a value of *HRT* ranging from the precipitation
435 value associated to the 20-th percentile of the PDC_{nz} (minimum *HRT*) to the value associated
436 to the 0.01-th percentile (maximum *HRT*).

437 All the results of the calibration procedure for each river basin, together with the climatic
438 variables representative of each calibration period, are synthesized in Table 4. The calibration
439 procedure has led to model parameters whose values are coherent with the values found in
440 Pumo et al. (2014). Only the parameter m_0 for the case of the *Nocella at Zucco* river basin
441 results significantly different from those computed for the other basins; this is due to the fact
442 that the *Nocella* basin exhibits an hydrological behavior rather different from the other basins,
443 being it characterized by a higher precipitation and a lower temperature, and, consequently, a
444 less ephemeral response in term of streamflow. Nevertheless, the resulting values of m_0 for
445 this site is consistent with the analogue value ($=21 \cdot 10^8$) found by Croker et al. (2003), where a
446 different formulation (i.e. only related to the *MAP*) to estimate p_{dry} at the regional level was

447 used; in fact, the application of the formulation by Croker et al. (2003), calibrated for
448 Portuguese catchments, to the case of *Nocella at Zucco* river basin would provide a value of
449 p_{dry} for the calibration period practically coincident with the value obtained by Eq. (1) with
450 the parameters provided in Table 4.

451 **2.5 Weather generator and downscaling procedure**

452 The future climatic scenarios here analyzed are generated by the hourly *Advanced*
453 *WEather GENerator*, AWE-GEN, (Ivanov et al 2007; Fatichi et al., 2011) adopting an
454 opportune downscaling procedure based on a stochastic downscaling of GCMs realizations
455 (Tebaldi et al., 2004 and 2005; Fatichi et al., 2011 and 2013). Since the hydrological model
456 calibration has been carried out using a database (i.e., OA-ARRA dataset) different from that
457 used to train the weather generator (i.e., SIAS dataset) and considering at each site a past
458 period not representative of the current conditions, the selected calibration periods are
459 scarcely suitable as reference scenarios to evaluate potential future changes in the hydro-
460 climatic variables. To this aim, a *Baseline* scenario, representative of the current conditions,
461 has been opportunely generated for each basin by the same AWE-GEN.

462 The stochastic downscaling method uses a Bayesian approach to weight climate model
463 realizations (Tebaldi et al., 2004; 2005; Fatichi et al., 2011), which allows the derivation of
464 the probability distributions of factors of change (FOCs) representative of an ensemble of
465 GCM projections. FOCs from single climate models can be calculated as “ratios” for
466 precipitation and “delta” for temperature (i.e., ratios and differences between climate statistics
467 of GCMs historical and future scenarios). More specifically, a set of factors of change is
468 computed to adjust mean monthly air temperature and several statistics of precipitation (e.g.,
469 mean, variance, skewness, frequency of no-precipitation) at different aggregation periods (24,
470 48, 72, 96 h), as a result of comparing historical and projected climate model outputs. The
471 factors of change derived from the ensemble of GCM realizations are successively used to
472 obtain statistics representative of future climate scenarios. Using such statistical properties, an
473 updated set of AWE-GEN parameters can be estimated. Each parameters set is calculated
474 assuming stationary climate for any considered period. Finally, the re-parameterized weather
475 generator is used to simulate hourly time series of climatic variables that are considered to be
476 representative of the future climate. For a more detailed description of the model and
477 procedures, see original papers by Ivanov et al. (2007) and Fatichi et al. (2011 and 2013), and

478 the model guidelines, publically available at the website: [http://www-](http://www-personal.umich.edu/~ivanov/HYDROWIT/Models.html)
479 [personal.umich.edu/~ivanov/HYDROWIT/Models.html](http://www-personal.umich.edu/~ivanov/HYDROWIT/Models.html).

480 In this study, realizations from a subset of thirty two GCMs, used by the IPCC-5AR (Moss
481 et al., 2010; IPCC 2013), have been considered. The IPCC-AR5 analyzed four greenhouse
482 gas concentration trajectories (*Representative Concentration Pathways: RCPs*), corresponding
483 to four possible hypotheses about the rise of greenhouse gases emission in the years to come.
484 The four RCPs (i.e., RCP 2.6, RCP 4.5, RCP 6, and RCP 8.5) are related to the possible
485 radiative values increase in the year 2100 with respect to the pre-industrial values (+2.6, +4.5,
486 +6.0, and +8.5 W/m², respectively). The GCM realizations corresponding to a “moderately
487 optimistic”, i.e. RCP 4.5, and the most “pessimistic”, i.e. RCP 8.5, IPCC-AR5 greenhouse gas
488 concentration trajectory have been considered in this study.

489 The GCMs outputs exhibit a large spread, underlining inherent uncertainties in climate
490 model predictions. This is particularly evident for precipitation where factors of change are
491 substantially different among the models, while the discrepancies relative to the detected
492 changes in air temperature results generally more restrained.

493 The method is here applied starting from hourly data from 2003 to 2013 derived from the
494 SIAS meteorological stations selected as representative of the five investigated case studies
495 (i.e., one station for each basin, see Table 3). Five climate scenarios have been created and
496 investigated for each selected case study: namely the *Baseline* and two climate projections, at
497 2055 and 2090, for both the RCPs 4.5 and 8.5, based on GCM projections over the periods
498 2045-2065 and 2081-2100, respectively. The downscaling procedure uses the GCMs outputs
499 of the grid cells containing the SIAS weather station selected as representative for each study
500 basin. Rainfall, temperature, solar radiation, relative humidity and wind speed have been
501 simulated in order to obtain stationary 50 years long time series for each climatic scenario.
502 For the generation at each site of the various scenarios, the weather generator has been re-
503 parameterized using only the median of each FOC, creating just one realization of the GCMs
504 ensemble. Only FOCs for precipitation and air temperature have been evaluated; therefore,
505 changes of the other required variables, i.e., solar radiation, relative humidity and wind speed,
506 are not a direct consequence of the calculated factors of change, but are only due to the
507 internal relationships embedded in the weather generator. The FOCs have been used to re-
508 estimate climate statistics for the central years (2055 and 2090) of the periods 2046-2065 and
509 2081-2100, while the *Baseline* scenario, assumed as the control scenario period for which

510 GCMs data and SIAS data are available, has been generated considering the same statistical
511 properties derived from the observation period (i.e., 2003-2013).

512 **3 Results and discussion**

513 **3.1 Hydrological model performances**

514 The ModABa performances are here evaluated at each basin in terms of ability in
515 reproducing the main characteristics of the hydrological regime during the calibration period.
516 Empirical and theoretical FDCs relative to the calibration period are compared evaluating the
517 distance index Δ , and it is also shown a comparison between some results arising from the
518 *Module Statistics* and the corresponding data empirically computed.

519 Figure 4 shows the comparison between model (black curves) and empirical (red points)
520 FDCs for the five basins, also reporting the associated value of Δ . Each curve, represented in
521 a semi-log plot, describes the various durations (days) associated to different specific
522 streamflow values (mm/day). Despite the five considered basins exhibit marked differences in
523 the main topographic characteristics (Table 1), in vegetation and soil covers (Table 2) and in
524 the main climatic forcings (Table 4), model performances in terms of Δ are comparable for all
525 the cases and consistent with the performances exhibited for the case of the *Eleuterio at Lupo*
526 in Pumo et al. (2014). For two basins (i.e., *Forgia at Lentina* and *Baiata at Sapone*) the
527 distance indexes are significantly lower than that shown for the case of *Eleuterio at Lupo* (i.e.,
528 $\Delta=0.0089$) denoting a better reproduction of the empirical FDC, and only a case (i.e., *Nocella*
529 *at Zucco*) has provided a slight higher value of Δ (i.e., $\Delta=0.0129$).

530 Most of the evidences arising from the previous application by Pumo et al. (2014) have
531 been here confirmed; for instance, model reproduction of the lower and more frequent
532 streamflow values has been extremely accurate for all the investigated cases, with near
533 coincident lower tails for the model and the corresponding empirical FDCs, while, in some
534 cases, a slight underestimation of the frequencies can be noticed in the part of the FDCs
535 representing the transient zone in which subsurface contribution becomes prevalent with
536 respect to the surface contribution (i.e., durations from about 3 to 9 days). Moreover, the
537 model often shows some difficulties in reproducing streamflow values relative to extremely
538 rare event, i.e. associated frequencies in the FDC corresponding to sub-daily durations. This
539 general behavior is highlighted in the example reported in Figure 5, relative to the case of the
540 *Forgia at Lentina*, where specific streamflow values are not log-transformed and the duration

541 domain is separated in two parts. In the same figure, the representation of also the SSFDC and
542 SFDC highlights the prevalence of the subsurface contribution to streamflow for the middle
543 and lower part of the FDC (right panel), while the surface contribution essentially controls the
544 upper tail (left panel).

545 Table 5 shows a comparison between some descriptors (i.e., values of the discharges for
546 fixed percent durations, mean discharge over the year and over the *non-zero season*, mean
547 durations of the *non-zero season*) of the hydrological regime simulated by the ModABa and
548 observed for the same period. Also such results confirm the general good accuracy of the
549 model in reproducing the observed hydrological behaviors of the investigated basins during
550 the calibration period. The model reproduces the discharge of various magnitudes with
551 acceptable discrepancies. Also in terms of mean discharge over the year (Q_m) and the *non-*
552 *zero season* ($Q_{m,nz}$), the differences between theoretical and empirical values are rather
553 moderate with an exception for the *Nocella at Zucco* river basins, where a significative
554 overestimation of about 120 l/s for both the values can be noticed. This outcome is affected
555 by a consistent overestimation of the streamflow values associated to very rare events
556 (durations lower than 7 days). Model reproduction of the *non-zero season* mean duration is
557 extremely accurate for the case of *Fastaia at La Chinaea*, while, for the other considered
558 basins, the ModABa provides a systematic overestimation on the order of 20 days (on
559 average, about the 10% of the season). This could be also due to the fact that the model
560 simulates also very low discharges, i.e. lower than streamflow gauge accuracy.

561 **3.2 Projected changes in precipitation and temperature**

562 Predicted climatic scenarios are similar in all the selected basins (Table 6), with a decrease
563 in precipitation and a simultaneous increase in temperature in the order of about the 10% for
564 both the considered temporal horizons under the RCP 4.5 and for the 2055 scenario under the
565 RCP 8.5, and about twofold percent variations for the 2090 under the RCP 8.5. Rainfall
566 events are projected to be, on average, less frequent and more intense over the year. The mean
567 annual precipitation percent reduction from *Baseline* to 2090 ranges, among the five basins,
568 from the 10.9% to the 12.4%, under the RCP 4.5, and from the 21.6% to the 23.8%, under the
569 RCP 8.5, while the temperature percent increment ranges from the 9.7% to the 12.4% under
570 the RCP 4.5, and from the 20.9% to the 25.3% under the RCP 8.5.

571 For instance, in the case of *Senore at Finocchiara*, the mean annual precipitation
572 decreases, under the RCP 4.5, from 659 mm, in *Baseline*, to 598 mm in 2055 and 583 mm in

573 2090 (about -76 mm in a century) while the mean annual temperature increases from 16.7 °C
574 to 18.2 °C in 2055 and 18.8 °C in 2090 (i.e. about +2°C in a century). Analogous and more
575 marked trends can be noticed for the same basin under the RCP 8.5, corresponding to a
576 greater increase in greenhouse gas concentration: in this case, the mean annual precipitation
577 decreases down to 587 mm in 2055 and 503 mm in 2090 (i.e. -156 mm), while the mean
578 temperature increases up to 18.9°C in 2055 and 20.8 °C in 2090 (i.e. about +4 °C).

579 A representation of the projected climatic changes at the monthly level is provided in
580 Figure 6 for the same case (*Senore at Finocchiara*), which shows, for each scenario (different
581 colors) generated by the AWE-GEN, the median and the 10th-90th percentile values of the
582 monthly precipitation P (Figures 6a and 6b) and temperature T (Figures 6c and 6d) under both
583 the RCPs 4.5 (left panels) and 8.5 (right panels). With respect to the *Baseline* scenario,
584 significant decreases in monthly precipitation are predicted under the RCP 4.5 (Figures 8a and
585 8c) especially for the winter months (about -44 mm in precipitation from December to March
586 in 2090), while the higher increases in temperature are mainly predicted during the summer
587 months (e.g., +2.7°C in August for the 2090 versus a minimum monthly increase of +1.6 °C
588 in February). The same behavior is projected also under the RCP 8.5 (Figures 6b and 6d),
589 with the most relevant decrements in monthly precipitation during wintry months (e.g., -25
590 mm in December for the 2090 scenario), and maximum temperature increments during the
591 summer (e.g., almost +5 °C in July and August in 2090 scenario).

592 **3.3 Projected changes in evapotranspiration**

593 Climate change may alter the entire energy balance at the basin scale, inducing
594 significant changes in the evapotranspiration processes. The increase in the air concentration
595 of CO₂, on the one hand, induces a temperature increase, which implies an increment also in
596 the atmospheric evaporative demand, while, on the other hand, it is expected to induce a
597 reduction in plants stomatal conductance to water vapor (Long et al., 2004, Moratiel et al.,
598 2011). Alterations of the leaf stomatal conductance with increasing CO₂, could potentially
599 reduce the energy used in evaporating processes, also modifying the mechanism of stomatal
600 regulation. This anticipated phenomenon could partially counterbalance the effects on
601 evapotranspiration induced by the temperature increment.

602 The coupled effect of climatic forcings and vegetation response is here taken into account
603 through the evaluation of the current and the future reference evapotranspiration series by the
604 FAO-56 Penman-Monteith formulation (Allen et al., 1998) and hypothesizing an opportune

605 function relating the stomatal conductance and the air CO₂ concentration. In particular,
606 starting from the estimation of the stomatal conductance of the standard vegetation associated
607 to the current atmospheric CO₂ concentration (i.e., 390 ppm), we referred to the results of
608 Long et al. (2004), which observed decreased stomatal conductance by about the 20%, on
609 average, for C3 plants grown in elevated CO₂ concentration (about 550 ppm) in *FACE (Free-*
610 *Air CO₂ Enrichment)* experiments, based on more than 200 independent measurements.

611 Given the overall uncertainty of the CO₂ effects in reducing stomatal conductance,
612 especially in the long-term when plants might undergo acclimation effects (Ainsworth and
613 Long, 2005), it has been here assumed a stepwise formulation, the same adopted in
614 Caracciolo et al. (2014), assuming a linear decrease of stomatal conductance from the current
615 value to a value reduced by the 20%, in accordance with Longo et al. (2004), for CO₂
616 concentration increasing from the current value to 550 ppm and assuming a constant stomatal
617 conductance for higher values of carbon dioxide concentration.

618 More specifically, the current stomatal conductance, referred to a hypothetical well-
619 watered grass with standardized characteristics (canopy height of 0.12 m, bulk surface
620 resistance of 70 s/m, albedo equal to 0.23, LAI_{max} equal to 2.88 m²·m⁻²) and equal to 0.0144
621 m/s according to Allen et al. (1998), has been assumed for the *Baseline* scenarios. According
622 to the projections of the IPCC (2013), the expected CO₂ concentrations for the projections
623 2055 are 535 ppm and 675 ppm under the RCPs 4.5 and 8.5, respectively, while for the
624 projections 2090 are 575 ppm and 1090 ppm under the RCPs 4.5 and 8.5, respectively. Thus,
625 for the scenarios 2055 of the RCP 4.5 the estimated stomatal conductance was 0.0118 m/s,
626 while, for all the other future scenarios, it was assumed the value of 0.0115 m/s (i.e., -20%,
627 corresponding to atmospheric CO₂ concentrations \geq 550 ppm).

628 A daily reference evapotranspiration series has been finally derived by FAO-56 Penman-
629 Monteith equation for each considered basin and climatic scenario, using, after daily
630 aggregation, the climatic series generated by the AWE-GEN for all the required climatic data
631 (i.e., temperature, solar radiation, wind speed, air relative humidity and air pressure).

632 The reference evapotranspiration values at the averaged monthly time-scale, ET_0 , obtained
633 for the case of *Senore at Finocchiara* are also represented in Figures 6e and 6f, while the
634 mean annual values for each basin are reported in Table 6. The monthly increments of
635 reference evapotranspiration follow those in temperature, with a maximum in summer months
636 (July and August) and a minimum in February. The variations predicted under the RCP 8.5
637 scenario are much more relevant than those relative to the RCP 4.5, especially with regards to

638 the summer months. At the annual level, it can be noticed how the induced percent variations
639 in the reference evapotranspiration values are rather dissimilar among the various basins,
640 differently from what observed with regard to the changes in both temperature and
641 precipitation. The projected reference evapotranspiration increments range from about the
642 0.35% (i.e., *Fastaia at La Chinea, 2055, RCP 4.5*) to the 8.2% (i.e., *Fastaia at La Chinea,*
643 *2090, RCP 8.5*). Finally, it is worth emphasizing that reference evapotranspiration have been
644 here obtained implicitly neglecting possible non-stationarity in solar net radiation, wind speed
645 and vapor pressure deficit time series.

646 **3.4 Projected changes in the hydrological regime**

647 The hydrological model has been forced by the daily series of precipitation and reference
648 evapotranspiration generated for each considered climatic scenarios, using the calibration
649 parameters given in Table 4. The five climatic scenarios generated for each basin (i.e.,
650 *Baseline*, and the future scenarios *2050* and *2090* for both the RCPs 4.5 and 8.5) account for
651 alterations induced by climate change on precipitation, temperature, and stomatal
652 conductance. Although the consideration of time invariant soil and vegetation patterns is
653 barely realistic, no hypotheses have been made about changes in land use during all the
654 analyzed temporal horizons, focusing in this manner on the potential alterations in the
655 hydrological regime only due to the predictions of the climate change models.

656 Prospective shifts in the hydrological regime of each basin are here investigated through
657 the analysis and the comparison of the FDCs and the different data provided by the *Module*
658 *Statistics* for each climatic scenario. The shape of the FDC, especially in its upper and lower
659 parts, could be particularly significant in characterizing the type of flow regime and the basin
660 ability to sustain low flows during the dry seasons. The evaluation of the mean annual
661 streamflow is directly related to the water resource availability assessment, while the analysis
662 of potential variations in the mean duration of the flow (*non-zero*) season could suggest
663 different strategies towards a more rational and sustainable use of the rivers and the
664 underlying ecosystems.

665 **3.4.1 The Baseline scenarios**

666 A first analysis consists in a qualitative comparison, for each basin, between the FDC
667 relative to the current *Baseline* scenario and the empirical FDC relative to the calibration
668 period (Figure 4). An Italian classification, according to the Legislative Decree n.131/2008,

669 distinguishes the non perennial rivers into three different classes: 1- *intermittent*, with flow
670 conditions during more than 8 months per year; 2- *ephemeral rivers*, with flow conditions
671 during less than 8 months per year; 3- *episodic*, usually dry and with flow conditions after
672 intense rainfall events. According to this classification and analyzing the empirical FDCs for
673 the calibration periods, all the selected case studies were in recent past belonging to the
674 *ephemeral rivers* class, with the exception of the *Nocella at Zucco*, where the past
675 hydrological regime can be classified as *intermittent*. The same classification results
676 maintained for all the basins also for the current scenarios, reproduced by the ModABa.

677 In all the cases, precipitation, temperature and evapotranspiration regime estimated for the
678 *Baseline* scenarios results rather similar to that observed during the calibration period: slight
679 differences can be mainly attributable to the different datasets used (OA-ARRA database in
680 calibration and SIAS database for the *Baseline* scenarios generation).

681 The model FDCs relative to the various climatic scenarios are shown in Figure 7. As a
682 consequence of the rather moderate difference in climatic forcings, the FDCs relative to the
683 *Baseline* scenarios (black curves) are almost coincident with the empirical FDCs for each
684 basin (red points in Figure 4). The most evident discrepancies between current-simulated and
685 past-observed FDCs can be noticed for the case of *Forgia at Lentina*, which, among the five
686 investigated basins, is that characterized by the highest distance between calibration and
687 *Baseline* climate; for this case, the *Baseline* FDC denotes a significative reduction of
688 streamflow over the entire durations domain with respect to the calibration period (i.e., 1971-
689 1985), while the shape of the curve remains essentially unaltered.

690 3.4.2 Projections under the RCP 4.5

691 The first considered emission scenario, namely the RCP 4.5, considers a greenhouse gas
692 concentration trajectory with a relevant initial increase of CO₂ up to about the 2060 and then a
693 sort of “plateau” for successive years. This obviously implies relevant changes on all the
694 hydro-climatic variables passing from the *Baseline* to the 2055 projection, and significantly
695 smoothed alterations passing from the scenario 2055 to the 2090. All this is actually here
696 confirmed by the analysis of both the climatic forcing and the resulting FDCs for all the case
697 studies. Dashed and solid gray curves in Figure 7 show the projected FDCs in 2055 and 2090,
698 respectively, under the RCP 4.5. It can be noticed how the curves approximately preserve the
699 *Baseline* FDCs shape, resulting significantly downshifted with respect to the *Baseline*,
700 especially in the middle-low part, while the upper tails of the FDCs result approximately

701 unaltered. The differences between the 2055 and 2090 FDCs are much less marked than those
702 between the *Baseline* and 2055 FDCs; for the case of *Baiata at Sapone* the two future FDCs
703 are practically coincident, while for the case of *Fastaia at La Chinaea*, the 2055 curve is even
704 lower (i.e., reduced streamflow) than that relative to the 2090. This is due to the fact that, for
705 both the cases, the two climatic projections (i.e. 2055 and 2090) are almost identical (Table
706 6), especially with regard to the only *non-zero season*, and the modifications induced in the
707 streamflow regime are essentially driven by changes in the characteristics of rainfall events,
708 that are projected more intense and less frequent in 2090.

709 The general behavior in terms of changes in the FDCs is also highlighted in Figure 8,
710 relative to two cases (i.e., *Nocella at Zucco* and *Senore at Finocchiaro*), where the specific
711 streamflow values are not log-transformed and the duration domain is represented in a
712 fragmented version, separately representing the highest and rarest streamflow values (i.e.,
713 with duration under 14 days), and the middle-low part of the FDC, with the most frequently
714 overcome streamflow values (i.e., durations over 14 days).

715 Under the RCP 4.5 scenario, the *Nocella at Zucco* basin has shown the lowest variations
716 among the five analyzed cases, with an almost unaltered hydrological regime in 2055 and a
717 slight reduction, for fixed duration, of almost all the streamflow values in 2090 projection. A
718 quantitative analysis can be performed also through the value reported in Table 7a. With
719 regard to the case of *Nocella at Zucco*, comparing the various discharges of equal percent
720 duration for the 2055 with the corresponding *Baseline* values, it can be noticed how higher
721 streamflow values for extremely rare events (with duration lower than about 4 days) partially
722 compensate for the reduction in discharges over almost all the years, keeping unaltered the
723 annual water availability of the river (i.e., almost constant mean annual runoff). This effect,
724 related to the modelling structure of the ModABa, also concerns the 2090 projection, where
725 the mean annual discharge is reduced by only the 6.4% despite a percent reduction in
726 precipitation and a percent increment in temperature of almost the 13%. In this basin,
727 moreover, both the mean duration of the *non-zero season* and the annual runoff coefficient
728 remain almost unchanged.

729 The second case reported in Figure 8 (i.e., the *Senore at Finocchiaro*) can be assumed as
730 representative of the behavior of all the other basins. With respect to the previously discussed
731 case, the modifications on the FDCs induced by climate change under the RCP 4.5 are here
732 much more relevant, especially for what concerns the middle-low part of the curves, while,
733 also in this case, the upper tails remain almost unaltered with higher streamflow values

734 associated to the lowest frequencies (i.e., durations lower than about 2 days). As it can be
735 noticed by Table 7a, the *non-zero season* of the *Senore at Finocchiara* basin is projected to be
736 significantly shorter in the future and, on average, characterized by a higher discharge (i.e.,
737 higher $Q_{m,nz}$). At the annual level, it can be observed a significative decrement of the available
738 water resource, with the mean annual runoff reduced about by the 10% and the 20% for the
739 projections at 2055 and 2090, respectively.

740 3.4.3 Projections under the RCP 8.5

741 The RCP 8.5 considers an exponential increase of the atmospheric CO₂ with a projected
742 concentration in 2100 over three times the current value. In all the basins here examined, the
743 variations from the *Baseline* to the 2055 in both precipitation and temperature under the RCP
744 8.5 results only slightly higher than under the RCP 4.5, while much more marked differences
745 can be noticed comparing the 2090 climatic scenarios of the two different concentration
746 trajectories (Table 6). Moreover, the 2055 precipitation, temperature and reference
747 evapotranspiration under the RCP 8.5 result quite similar to the corresponding forcings
748 obtained for the scenario 2090 under the RCP 4.5.

749 Under the RCP 8.5, the percent variations in both precipitation and temperature from the
750 *Baseline* to the 2090 temporal horizon are about two times those projected passing from the
751 *Baseline* to the 2055, and even more accentuated differences can be noticed with regard to the
752 reference evapotranspiration. Such a behavior is also reflected in the hydrological response of
753 the various basins to the different climatic scenarios. From the observation of the FDCs
754 (Figures 7 and 8), one can notice that the FDCs projected in 2055 under the RCP 8.5 (red
755 dashed curves) are quite close to those relative to the 2090 under the RCP 4.5 (black solid
756 curves), while the FDCs projected in 2090 under the RCP 8.5 show a deep modification in the
757 hydrological regime of the analyzed basins; actually, in some of the basins previously
758 classified as *ephemerals*, the flow conditions in the future could mainly (or totally) rely on the
759 precipitation regime, approaching to hydrological conditions typical of *episodic* river basins.

760 Some basic statistics obtained for the RCP 8.5 and synthesized in Table 7b, project, on
761 average, a reduction of about the 13% in the annual water availability in 2055, while, for the
762 2090, the projected percent reduction results, on average, of about the 32%, with a maximum
763 of almost the 39% for *Baiata at Sapone*. A very important aspect is the relevant shortening of
764 the *non-zero season*, which in some cases (i.e., *Forgia at Lentina* and *Senore at Finocchiara*)
765 is projected to be almost halved for the 2090. Despite a mean annual runoff coefficient only

766 slightly reduced with respect to the *Baseline*, the hydrological regime for the 2090 projections
767 appear significantly modified. Although streamflow values in 2090 are projected to be lower
768 than the current values over almost all the year (i.e., for more than the 99% of the year), the
769 mean discharge over the *non-zero season* is projected to increase; this behavior, observed for
770 all the cases with the exception of the *Nocella at Zucco*, is again related to the hydrological
771 model structure and is the result of the projected higher occurrence of extreme streamflow
772 values with very low associated frequencies (i.e., durations in the order of 1 day or less),
773 probably due to heavy rain events that are projected to occur more frequently and with a
774 higher intensity in the future. Although this last aspect may reflect the uncertainties of GCMs
775 in reproducing rainfall extremes, it is worth mentioning that a good capability in reproducing
776 historical trends by GCMs was recently recognized (e.g., Tebaldi et al., 2006).

777 3.4.4 Projected changes in streamflow components

778 The adopted hydrological model allows for the evaluation of the relative importance of the
779 two streamflow components (i.e., surface and subsurface) in contributing to the runoff both at
780 the annual and the daily level. For each analyzed climatic scenario, the percent contribution of
781 the surface flow component (*SFC*) to the mean annual runoff (*R*) has been evaluated.
782 Unfortunately no direct observations are available for such a kind of analysis and, therefore,
783 the respective results, discussed in this section, should be interpreted with some degree of
784 uncertainty given the impossibility to perform a rigorous validation.

785 With regard to the *Baseline*, despite the daily discharge during the most part of the year is
786 mainly formed by the subsurface component, the surface contribution to the mean annual
787 runoff results prevalent in all the basins (Table 7), with the only exception of the *Fastaia at*
788 *La Chinea* river basin (*SFC*= 41.3%); this behavior actually reflects the typical torrential
789 nature of most of the Sicilian rivers. The *Baiata at Sapone* is the basin showing the highest
790 value of *SFC* (=77.6%), while for the other cases the surface component weights on the total
791 annual runoff for about the 60%.

792 According to our results (Table 7), potential effects of climate change on the analyzed
793 basins could entail also the repartition between the two streamflow components; for both the
794 considered temporal horizon (2055 and 2090), the relative weight of the surface component is
795 predicted to significantly increase, especially with regard to the RCP 8.5 scenario, further
796 enhancing the torrential character of the considered basins. For the case of the *Senore at*
797 *Finocchiarà*, the value of *SFC* could even increase from the 58% to almost the 86% in the

798 next 80 years, under the prospective of the RCP 8.5. The projected surface (*SFC*) and
799 subsurface (*SSFC*) percent contributions to the mean annual runoff, *R*, under the RCP 8.5 are
800 also represented in Figure 9, where it is also possible to compare the different value of *R*
801 provided by the five basins. It can be noticed that the *R* reduction over the two analyzed
802 temporal horizons, especially for the cases of *Fastaia at La Chinaea* and *Senore at*
803 *Finocchiara*, is mainly due to a reduction in the *SSFC* contribution, probably attributable to
804 heavier evapotranspiration processes, while the projected increases in *SFC* imply only a
805 moderate reduction of the surface component volumetric contribution to the mean annual flow
806 volume, mainly attributable to changes in the rainfall structure and seasonal distribution.

807 The discharge partitioning, at the daily level, between the surface flow component and the
808 subsurface flow component over the *non-zero season* is represented in Figure 10 for three
809 basins, which represents the cases with the highest and the lowest current value of *SFC*
810 (*Baiata at Sapone* and *Fastaia at La Chinaea*, respectively) and the case showing the highest
811 projected variations in terms of *SFC* (i.e. *Senore at Finocchiara*). The figure compares the
812 *Baseline* with the 2090 curves under both the RCPs 4.5 and 8.5. Each curve represents the
813 surface component percent contribution relative to the various streamflow values as a function
814 of the corresponding frequency (i.e. percent durations) in the FDC_{nz} ; thus, each point of the
815 curves provides the streamflow component repartition as a function of the percentage of the
816 *non-zero season* for which the corresponding total streamflow is exceeded. Only the portion
817 of the *non-zero season* where the surface contribution weights on the daily streamflow for
818 more than the 10%, corresponding to the streamflow values associated to the lower durations
819 (i.e., higher streamflow values), is represented.

820 Under the current conditions (i.e., *Baseline*) the percent contribution of surface component
821 at the daily level (SFC_i) is significative (>10%) for streamflow values that are exceeded over
822 about the 20% of the *non-zero season* at the *Baiata at Sapone* and *Fastaia at Finocchiara*
823 basins, and about the 24% at the *Senore at Finocchiara* basin. Significative future
824 modifications on this aspect result only for the case of *Senore at Finocchiara*, where the
825 portion of the *non-zero season* with streamflow values having $SFC_i > 10\%$ could increase up
826 to the 33%. The surface contribution is dominant ($SFC_i > 90\%$) for streamflow values that are
827 currently exceeded only for an extremely limited part of the flow season (less than the 2% for
828 all the cases) and no relevant future variations are projected on this aspect.

829 Substantial changes are, conversely, projected for the 2090 scenario under the RCP 8.5
830 with regard to the portion of the *non-zero season* characterized by streamflow values mainly

831 formed by the surface component. In fact, from Figure 10, it can be noticed how, passing
832 from the *Baseline* to 2090 scenario of the RCP 8.5, the percentage of the *non-zero season*
833 characterized by streamflow with $SFC_i > 50\%$ increases from the 4% to almost the 26% for the
834 case of *Senore at Finocchiara*, and from the 12% to the 17% for the *Baiata at Sapone* basin,
835 while it is almost unaltered for the case of *Fastaia at La Chinea*.

836 **4 Conclusions**

837 The analyses here performed has highlighted potential impact of climate change upon
838 freshwater resource in five small basins within the northwestern Sicily, investigating possible
839 modifications in the respective flow duration curve (FDC). In this paper, a recent hydrological
840 model, the ModABa, and a weather generator, the AWE-GEN, have been coupled to analyze
841 the effects of some generated future climate scenarios on the mean annual runoff and on the
842 daily streamflow amount and composition (in terms of repartition between surface and
843 subsurface components) for the selected case studies.

844 An opportune calibration procedure makes the adopted hydrological model particularly
845 suitable to work under variable climatic conditions. The model has been first calibrated at
846 each site with regard to a past “known” period, testing its performances that have been
847 revealed, for all the cases here investigated, coherent and comparable with the performances
848 previously shown in Pumo et al. (2014). For each basin, a *Baseline* scenario representative of
849 the current conditions, which maintains the statistical properties of recently (i.e., from 2003 to
850 2013) observed series of precipitation, wind speed, air temperature, relative humidity and
851 pressure, has been generated by the AWE-GEN. The future scenarios generated by the same
852 model refer to two different temporal horizons, in order to investigate potential changes
853 induced in about half century (projection at 2055) and in about a century (projection at 2090)
854 under two possible different greenhouse gases emission scenarios (i.e., RCPs 4.5 and 8.5 by
855 the IPCC, 2013).

856 Our findings have demonstrated that climate change is likely to have substantial effects on
857 the hydrological regime of typical non perennial small basins of the southern Mediterranean
858 region. For all the selected basins, all climate change scenarios project a significative
859 reduction in the mean annual discharge, which mainly reflects the changes in precipitation, as
860 it is demonstrated by the scarcely altered annual runoff coefficient. The results show a
861 different response to climate change across the basins here considered, in consideration of the
862 different geomorphologic characteristics, basin size, vegetational and soil patterns; the

863 strongest changes have been found for the basin of *Senore at Finocchiara*, which is the largest
864 considered basin, with the highest fraction of clayey soils and grass vegetation.

865 Beside the future water resources reduction, which is a common outcome for
866 Mediterranean case studies, the novelty of this work consists in the analysis of future
867 streamflow seasonality and distinction between surface and subsurface components. In fact,
868 large alterations to streamflow seasonality are projected so that in the future we may expect,
869 according to our predictions, more intense winter flood and more prolonged droughts in
870 summer. The season usually dry could be subject to a general lengthening, also in the order of
871 about three months in a century, under the most “pessimistic” emission scenario (RCP 8.5),
872 and the hydrologic regime could be shifted from the current conditions (classified as
873 *ephemeral*) towards conditions typical of *episodical* torrential rivers, where discharge totally
874 relies on precipitation. This prevision is also strengthened by the simulated projected changes
875 in the repartition between the two main components of streamflow, both at the annual and the
876 daily level. The relative importance of the surface component in contributing to the mean
877 annual runoff is projected to significantly increase. Flow conditions during the *non-zero*
878 *season* are projected to be characterized by higher streamflow values, on average, with respect
879 to the current conditions, with a more frequency of intensive floods due to more frequent and
880 intense heavy rains.

881 This study also demonstrates that the change induced in the hydrological response of the
882 basin is strongly related to the considered rate of growth in carbon dioxide concentrations in
883 the atmosphere; in fact, the projected changes in 2090 under the RCP 4.5 are comparable with
884 those predicted for the 2055 under the RCP 8.5, while the previsions for 2090 under the RCP
885 8.5 depict, for the analyzed basins, an alarming future scenario, with important and potentially
886 dramatic implications for the biotic component of the underlying ecosystems and for which
887 opportune counter-measures should be thought by local authorities.

888 Our outcomes, arising from an integrated approach linking climate, hydrological and
889 ecosystem models, could be affected by different sources of uncertainty mainly associated
890 with the hydrological model structure and parameters, with the observations used to drive and
891 evaluate the two models (ModABa and AWE-GEN), and with the climate scenarios
892 generation. Although an adequate ModABa reliability and accuracy in the reproduction of the
893 FDC and water availability has been proved by both the past (Pumo et al., 2014) and the five
894 here presented model applications to real cases, the assessment of model reliability in terms of
895 evaluation of the streamflow component repartition would need further analysis. Furthermore,

896 different techniques in the GCMs downscaling might give different estimates of the climatic
897 variables changes and, consequently, of the induced variations in the hydrological regime of
898 the investigated river basins. In spite of the uncertainties, common to many other studies on
899 this topic, this work may be considered as a benchmark for studies aimed to future water
900 resources assessment and could provide a sound basis for supporting decision making on
901 water resources management in the analyzed region.
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903 **References**

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1059 Table 1. Main features of the five case studies: A = Total area [km^2]; L = main river channel
1060 length [km]; H_m = mean elevation [m a.s.l.]; c_0 = fraction of impervious area [%].

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River Basin	A	L	H_m	c_0
	[km^2]	[km]	[m a.s.l.]	[%]
<i>Forgia at Lentina</i>	50.06	23.9	287	7.22
<i>Baiata at Sapone</i>	29.06	11.5	111	7.14
<i>Fastaia at La China</i>	23.94	7.1	327	8.35
<i>Nocella at Zucco</i>	61.57	14.7	527	4.29
<i>Senore at Finocchiaro</i>	75.92	23.6	407	7.37

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1078 Table 2. Parameters and relative percent cover relative to each vegetation (a) and soil (b) type
 1079 present within the permeable portion of the five study basins. Z_r = mean rooting depth; δ_{veg} =
 1080 canopy interception threshold; K_s = saturated hydraulic conductivity; β = coefficient of the
 1081 hydraulic conductivity power law; n = porosity; s_w = wilting point; s_{fc} = field capacity.

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(a)

VEGETATION	Parameters			Relative Percentage				
	Z_r	δ_{veg}	[%]					
	(mm)	(mm)	Forgia	Baiata	Fastaia	Nocella	Senore	
Trees	1000	2.0	63.6	34.8	51.3	5.9	29.9	
Shrubs	400	1.0	12.8	51.3	8.3	70.0	16.6	
Grasses	150	0.5	23.6	13.9	40.4	24.1	53.5	

(b)

SOIL	Parameters					Relative Percentage				
	K_s	β	n	s_w	s_{fc}	[%]				
	(mm/day)					Forgia	Baiata	Fastaia	Nocella	Senore
Sandy loam	607	13.8	0.415	0.22	0.47	7.0	-	15.7	46.9	-
silty loam	473	11.7	0.462	0.23	0.61	1.2	-	0.5	1.3	13.1
loam	227	14.2	0.465	0.26	0.55	15.6	52.8	12.7	0.7	1.4
silty clay loam	101	15.9	0.518	0.36	0.69	-	44.3	-	-	4.2
clay loam	90	16.3	0.504	0.35	0.65	31.6	2.9	-	4.1	20.0
Silty clay	65	19.2	0.532	0.47	0.77	22.7	-	25.0	36.9	-
Sandy clay loam	43	17.4	0.527	0.51	0.77	21.8	-	46.1	10.2	61.4

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1092 Table 3. Historical database by OA-ARRA used to derive historical series of daily
 1093 streamflow, rainfall and temperature for the hydrological model calibration and database by

1094 SIAS used to generate recent series of meteorological hourly data needed for the Weather
 1095 Generator. For each case study the name (with ID-code and coordinates), the available and the
 1096 considered dataset for both the hydrometric and the meteorological stations are listed. Name
 1097 and typology of each station by OA-ARRA considered to estimate areal rainfall and
 1098 temperature historical series at each basin are also reported (P = *Pluviometric station*; T =
 1099 *Thermometric station*).

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measured variables		<i>Forgia</i>	<i>Baiata</i>	<i>Fastaia</i>	<i>Nocella</i>	<i>Senore</i>	
Historical series (Hydrological Model) source: OA-ARRA	streamflow (daily)	Name	Forgia at Lentina	Baiata at Sapone	Fastaia at La China	Nocella at Zucco	Senore at Finocchiarra
		ID code	3710	3720	3730	3680	3795
		Coordinates	38.05N 12.67E	37.97N 12.60E	37.93N 12.75E	38.08N 13.11E	37.74N 13.03E
		av. dataset (size/from/to)	29 years (3 miss.) 1971 2002	27 years (7 miss.) 1968 2001	35 years (2 miss.) 1962 1998	42 years (4 miss.) 1958 2003	25 years (1 miss.) 1961 1986
	Selected dataset (Calibration Period) (size/from/to)	15 years 1971 1985	17 years 1974 1990	18 years 1967 1984	23 years 1977 1999	18 years 1969 1986	
Areal rainfall and temperature (daily)	Considered stations:	Erice (P); Lentina (P); Trapani (PT); Diga Paceco (PT); Calatafimi (P); Brogo Fazio (PT); Diga Rubino (PT); Fastaia (P); Specchia (P); Pioppo (PT); Partinico (PT); Montelepre (PT); Romitello (P); Montevago (P); Contessa Entellina (P); P.te Belice (PT); Vaccarizzo (P); Sambuca di Sicilia (PT); Giuliana (P); Chiusa Scalfani (P); Bisacquino (P); Diga Arancio (P); Corleone (PT); S.Margherita Belice (PT); S.Giuseppe Jato (PT); S.Giorgio C.C.(P); Vita (P); Segesta (P); Cuddia C.C. (P); Capo S.Vito (P); Dammusi (P); Carini (P); Salaparuta (P); Ginestra (P); Poggio S.Francesco (P)					
Recent series (Weather Generator) source: SIAS	rainfall, temperature, solar radiation, wind speed, air relative humidity and air pressure (hourly)	Name	Erice	Trapani Fulgatore	Calatafimi	Partinico	Contessa Entellina
		ID code	303	308	300	277	267
		Coordinates	38.03N 12.59E	37.95N 12.66E	37.85N 12.88E	38.07N 13.09E	37.72N 13.04E
	Selected dataset (size/from/to)	11 years from 2003 to 2013					

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1106 Table 4. Hydrological model calibration parameters and climate variables representative of
 1107 the calibration period for the five river basins. Fraction of impervious areas (c_0); mean soil

1108 and vegetation parameters (symbols are the same used for Table 2); regression parameters for
 1109 Eqs.(1) and (2); number and size of the subperiods considered in the regression analyses;
 1110 mean annual values of precipitation (MAP), rainfall events frequency (λ) and depth (α),
 1111 fraction of no-rainy days (f_{NR}), temperature (T) and daily reference evapotranspiration (ET_0).

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			<i>Forgia</i>	<i>Baiata</i>	<i>Fastaia</i>	<i>Nocella</i>	<i>Senore</i>	
impervious fraction	c_0	-	0.072	0.071	0.084	0.043	0.074	
mean vegetation parameters	Z_r	(mm)	722.5	574.2	606.7	375.3	445.5	
	δ_{veg}	(mm)	1.52	1.28	1.31	0.94	1.03	
mean soil parameters	K_s	(mm/day)	136.6	167.2	162.8	324.2	113.7	
	β	-	16.65	14.99	16.85	16.25	16.32	
	n	-	0.503	0.490	0.503	0.474	0.513	
	s_w	-	0.39	0.31	0.42	0.35	0.43	
	s_{fc}	-	0.67	0.62	0.69	0.62	0.72	
Regression parameters of Eq.(1)	<i>Subperiods for Eq.(1)</i>	<i>number</i>	10	15	45	45	21	
		<i>size (years)</i>	≥ 12	≥ 13	≥ 10	≥ 15	≥ 13	
	estimate of p_{dry}	m_0	(year/mm)	0.5	13.3	747.2	39270527	12.6
		m_1	-	-0.18	-0.41	-1.29	-2.91	-0.64
m_2		-	-3.20	2.08	-2.99	0.43	-1.95	
Regression parameters of Eq.(2)	<i>Subperiods for Eq.(2)</i>	<i>number</i>	4	4	5	9	5	
		<i>size (years)</i>	12	14	14	15	14	
	estimate of k	a_1	(1/day)	0.048	0.165	-0.321	0.173	0.113
		b_1	-	1.083	0.114	-5.519	0.207	0.663
		c_1	(1/mm)	-0.250	-0.139	2.106	-0.088	-0.232
	estimate of k_v	a_2	-	2.642	7.802	-1.740	-6.381	-0.641
b_2		(day)	0.396	2.762	22.364	8.440	-1.838	
c_2		(day/mm)	-0.958	-4.909	-3.894	2.693	2.781	
heavy rain threshold	HRT	(mm/day)	21.2	19.8	18.2	34.0	19.4	
Climate conditions for the calibration period	MAP	(mm/year)	674	650	686	974	654	
	λ	(1/day)	0.304	0.283	0.302	0.299	0.318	
	α	(mm)	6.10	6.29	6.21	8.93	5.62	
	f_{NR}	-	0.696	0.717	0.698	0.701	0.682	
	T	(°C)	18.1	18.7	17.6	17.3	16.4	
	ET_0	(mm/day)	2.43	2.52	2.38	2.36	2.24	

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1115 Table 5. Comparison between model and empirical values relative to the calibration period
 1116 for: various discharges (l/s) of equal percent duration; mean annual discharge (Q_m) and mean
 1117 discharge over the *non-zero season* ($Q_{m,nz}$) in l/s; mean durations of the *non-zero season* (d_{nz})
 1118 in days.

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		Discharge of equal percent duration (l/s)									
		Forgia		Baiata		Fastaia		Nocella		Senore	
		Emp.	Theo.	Emp.	Theo.	Emp.	Theo.	Emp.	Theo.	Emp.	Theo.
Percent duration	80%	0	0	0	0	0	0	10	17	0	0
	40%	30	35	10	11	5	7	120	128	40	40
	20%	170	176	30	38	77	69	290	295	180	198
	10%	410	383	80	72	150	163	550	619	490	454
	5%	750	766	180	203	325	357	950	978	1020	951
	1%	3650	1860	740	652	2120	923	2890	3662	5540	2763
Q_m (l/s)	202	201	42	71	101	97	243	365	302	315	
$Q_{m,nz}$ (l/s)	415	377	100	138	221	208	284	398	611	571	
d_{nz} (day)	178	195	152	189	166	169	312	334	180	201	

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1122 Table 6. Mean characteristics representative of the five climatic scenarios (current or *Baseline*
1123 scenario and future scenarios: *2055* and *2090* for both the *RCP 4.5* and *RCP 8.5*) generated by
1124 the AWE-GEN for each considered river basin. Symbols are the same used in Table 4.
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<i>river basin:</i>		<i>Forgia</i>					<i>Baiata</i>				
<i>scenario:</i>		<i>Baseline</i>	<i>RCP 4.5</i>		<i>RCP 8.5</i>		<i>Baseline</i>	<i>RCP 4.5</i>		<i>RCP 8.5</i>	
			<i>2055</i>	<i>2090</i>	<i>2055</i>	<i>2090</i>		<i>2055</i>	<i>2090</i>	<i>2055</i>	<i>2090</i>
<i>MAP</i>	(<i>mm/year</i>)	649	589	571	579	501	649	593	578	578	509
λ	(<i>1/day</i>)	0.293	0.234	0.228	0.221	0.186	0.277	0.252	0.229	0.230	0.194
α	(<i>mm</i>)	6.07	6.90	6.87	7.18	7.37	6.40	6.44	6.92	6.87	7.20
f_{NR}	-	0.707	0.766	0.772	0.779	0.814	0.723	0.748	0.771	0.770	0.806
<i>T</i>	($^{\circ}\text{C}$)	17.8	19.4	20.0	20.0	21.9	19.1	20.6	20.9	21.0	23.1
ET_0	(<i>mm/day</i>)	2.40	2.45	2.47	2.46	2.56	2.55	2.59	2.59	2.59	2.69
<i>river basin:</i>		<i>Fastaia</i>					<i>Nocella</i>				
<i>scenario:</i>		<i>Baseline</i>	<i>RCP 4.5</i>		<i>RCP 8.5</i>		<i>Baseline</i>	<i>RCP 4.5</i>		<i>RCP 8.5</i>	
			<i>2055</i>	<i>2090</i>	<i>2055</i>	<i>2090</i>		<i>2055</i>	<i>2090</i>	<i>2055</i>	<i>2090</i>
<i>MAP</i>	(<i>mm/year</i>)	664	606	590	593	517	944	850	827	837	719
λ	(<i>1/day</i>)	0.336	0.287	0.275	0.275	0.231	0.333	0.267	0.269	0.268	0.232
α	(<i>mm</i>)	5.42	5.80	5.88	5.90	6.13	7.76	8.72	8.44	8.55	8.48
f_{NR}	-	0.664	0.713	0.725	0.725	0.769	0.667	0.733	0.731	0.732	0.768
<i>T</i>	($^{\circ}\text{C}$)	17.3	19.0	19.3	19.6	21.3	16.8	18.3	18.8	19.1	21.0
ET_0	(<i>mm/day</i>)	2.40	2.41	2.41	2.43	2.59	2.33	2.36	2.38	2.39	2.48
<i>river basin:</i>		<i>Senore</i>									
<i>scenario:</i>		<i>Baseline</i>	<i>RCP 4.5</i>		<i>RCP 8.5</i>						
			<i>2055</i>	<i>2090</i>	<i>2055</i>	<i>2090</i>					
<i>MAP</i>	(<i>mm/year</i>)	659	598	583	587	503					
λ	(<i>1/day</i>)	0.303	0.250	0.245	0.256	0.214					
α	(<i>mm</i>)	5.95	6.54	6.51	6.28	6.46					
f_{NR}	-	0.697	0.750	0.755	0.744	0.786					
<i>T</i>	($^{\circ}\text{C}$)	16.7	18.2	18.8	18.9	20.8					
ET_0	(<i>mm/day</i>)	2.27	2.28	2.30	2.31	2.41					

1126

1127 Table 7. Results from *Module Statistics* under the RCP 4.5 (a) and the RCP 8.5 (b).
 1128 Discharges (l/s) of various percent duration; mean annual discharge (Q_m) and mean discharge
 1129 over the *non-zero season* ($Q_{m,nz}$) in l/s; mean durations of the *non-zero season* (d_{nz}) in days;
 1130 mean annual runoff coefficient (ϕ); mean annual runoff in mm/yr (R); mean percent
 1131 contribution of the surface component to the annual runoff (SFC).

1132

a) **Discharge of equal percent duration (l/s)**

RCP 4.5	Forgia			Baiata			Fastaia			Nocella			Senore			
	Baseline	2055	2090	Baseline	2055	2090	Baseline	2055	2090	Baseline	2055	2090	Baseline	2055	2090	
Percent duration	80%	0	0	0	0	0	0	0	0	9	9	5	0	0	0	
	40%	12	4	0	7	2	2	2	0	0	101	98	77	33	9	3
	20%	92	58	34	30	15	15	57	21	27	285	236	201	166	91	53
	10%	270	193	142	67	39	33	130	69	100	575	502	457	461	303	228
	5%	585	506	436	186	144	134	273	197	263	903	838	753	918	742	623
	1%	1562	1534	1400	597	506	531	713	617	799	1750	2263	1672	2251	2316	1934
Q_m (l/s)	167	151	131	75	55	61	64	53	63	304	303	285	252	227	203	
$Q_{m,nz}$ (l/s)	334	340	335	150	130	143	156	145	175	340	343	328	451	457	463	
d_{nz} (day)	183	162	142	182	156	156	150	132	132	327	322	317	204	181	160	
ϕ (-)	0.16	0.16	0.14	0.12	0.10	0.11	0.13	0.11	0.14	0.17	0.18	0.18	0.16	0.16	0.14	
R (mm/yr)	105	95	82	81	60	66	84	69	83	156	155	146	105	94	84	
MAC_{sc} (%)	60.1	64.9	68.8	77.6	82.4	83.5	41.3	59.7	49.2	57.5	59.6	64.0	58.0	68.0	73.6	

b) **Discharge of equal percent duration (l/s)**

RCP 8.5	Forgia			Baiata			Fastaia			Nocella			Senore			
	Baseline	2055	2090	Baseline	2055	2090	Baseline	2055	2090	Baseline	2055	2090	Baseline	2055	2090	
Percent duration	80%	0	0	0	0	0	0	0	0	9	7	0	0	0	0	
	40%	12	0	0	7	0	0	2	0	0	101	89	30	33	3	0
	20%	92	36	5	30	12	4	57	27	9	285	219	101	166	54	6
	10%	270	145	44	67	30	14	130	96	55	575	480	309	461	235	80
	5%	585	441	282	186	130	94	273	259	190	903	805	600	918	633	401
	1%	1562	1459	1214	597	538	465	713	789	703	1750	1709	1454	2251	1963	1595
Q_m (l/s)	167	137	103	75	61	46	64	63	50	304	270	216	252	205	167	
$Q_{m,nz}$ (l/s)	334	346	407	150	154	163	156	175	173	340	308	285	451	465	632	
d_{nz} (day)	183	144	93	182	145	102	150	132	106	327	321	276	204	161	97	
ϕ (-)	0.16	0.15	0.13	0.12	0.12	0.10	0.13	0.14	0.13	0.17	0.17	0.15	0.16	0.14	0.14	
R (mm/yr)	105	86	65	81	67	49	84	83	66	156	139	110	105	85	70	
MAC_{sc} (%)	60.1	69.0	78.7	77.6	86.1	89.3	41.3	49.7	56.6	57.5	58.3	72.9	58.0	73.8	85.7	

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