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# 22 MODELLING EFFECTS OF FOREST FIRE AND POST-FIRE MANAGEMENT IN A

### 23 CATCHMENT PRONE TO EROSION: Impacts on sediment yield

De Girolamo Anna Maria<sup>(a)\*</sup>, Cerdan Olivier<sup>(b)</sup>, Grangeon Thomas<sup>(b)</sup>, Ricci Giovanni Francesco<sup>(c)</sup>,
Vandromme Rosalie<sup>(b)</sup>, Lo Porto Antonio<sup>(a)</sup>

26 (a) Water Research Institute, National Research Council, Bari, Italy

27 (b) Bureau de Recherches Géologiques et Minières, Département Risques et Prévention, Orléans, France

28 (c) University of Bari Aldo Moro, Department of Agricultural and Environmental Sciences, Bari, Italy

29 \* Corresponding author: annamaria.degirolamo@ba.irsa.cnr.it

#### 30 Abstract

The present paper contributes to bridging the gaps in modelling post-fire impact and mitigation measures on 31 soil erosion. The specific aims were to predict the effects of forest fires and post-fire mitigation measures on 32 33 runoff and specific sediment yield (SSY) in a river basin (Celone, S-E Italy). The Soil and Water Assessment Tool model, calibrated with field observations, was used to evaluate runoff and SSY for the current land use 34 (baseline) and six post-fire scenarios. From 1990 to 2011, at the basin scale, the average annual SSY was 35 5.60 t ha<sup>-1</sup> vr<sup>-1</sup> (SD = 3.47 t ha<sup>-1</sup> vr<sup>-1</sup>). 20% of the total drainage area showed a critical value of SSY (>10 t 36 ha<sup>-1</sup> yr<sup>-1</sup>). The effects of different fire-severity levels were predicted for one year after the fire, acting on a 37 limited area (2.3% of the total basin area). At the basin scale, the post-fire effect on surface runoff was 38 negligible for all scenarios (< 0.4%), and the impact on SSY increased from 5.86 t ha<sup>-1</sup> yr<sup>-1</sup> up to 12.05 t ha<sup>-1</sup> 39 vr<sup>-1</sup>. At the subbasin scale, the post-fire logging scenario showed the highest increase of soil loss (SSY 40 increased from 9.48 t ha<sup>-1</sup> yr<sup>-1</sup> to 57.40 t ha<sup>-1</sup> yr<sup>-1</sup>). Post-fire mitigation treatments like straw mulching and 41 erosion barriers effectively reduced soil erosion in high- and moderate-severity fires (19.12 t ha<sup>-1</sup>yr<sup>-1</sup> and 42 20.93 t ha<sup>-1</sup> yr<sup>-1</sup>, respectively). At the hydrological response unit level, the SSY estimated for the forest in the 43 baseline ranged from 1.18 t ha<sup>-1</sup>yr<sup>-1</sup> to 2.04 t ha<sup>-1</sup> yr<sup>-1</sup>. SSY increased more than one order of magnitude for 44 the high-severity fire scenarios and ranged from 4.33 to 6.74 t  $ha^{-1}$  yr<sup>-1</sup> in the very low-severity fire scenario. 45 underlining the scale effect from the HRU to the basin scale. 46

47 *Keywords*: forest fires, sediment yield, runoff, SWAT model, fire severity, post-fire mitigation measures,
48 model parametrisation

### 49 **1. Introduction**

The Mediterranean European Region is a high fire risk area due to a combination of several factors. The high 50 51 number of buildings has increased the probability of fire ignition by human causes (Ganteaume et al., 2013) and, the abandonment of some rural areas has led to an accumulation of fuel loads that have contributed to 52 53 fire ignition and spread, especially in summer (San-Miguel-Ayanz et al., 2012). In addition to natural causes, 54 negligent and deliberate behaviors have contributed to increasing wildfires. Consequently, many fire events 55 are recorded every year in this region (Fernandéz-Anez et al., 2021). San-Miguel-Ayanz et al. (2012) 56 estimated that, in Europe, around 65000 fire events occur every year, burning about half a million hectares of 57 forest. The European Environmental Agency (European Commission, 2019) pointed out that the 58 meteorological fire hazard has increased due to climate change. The last report on forest fires (European 59 Commission, 2021; San-Miguel-Ayanz et al., 2021) highlighted in the recent years (2019, 2020, 2021) a 60 positive trend of higher levels of fire danger in EU countries, despite the increased level of preparedness. 61 About 340000 ha in 2020 and 0.5 million of ha in 2021 were burnt in the EU. About 25% of the burnt areas 62 were inside the "Natura 2000" sites. Several researchers pointed out that droughts and high temperatures 63 promote large fires in southern Europe (Camia and Amatulli, 2009; Lasaponara et al., 2018; Urbieta et al., 64 2015) that are also related to antecedent climate variables (Ruffault et al., 2016). Turco et al. (2017) highlighted that rising temperatures and droughts, which greatly influence summer fires, could make all fire 65 66 prevention efforts useless in the next decades.

Wildfires may result in serious economic, cultural, and ecological damages in the Mediterranean Region (Ganteaume et al., 2021). A forest fire is a disturbance for the ecosystem; it may alter soil properties (Mataix-Solera et al., 2011; Lucas-Borja et al., 2018), reduce infiltration capacity and increase the peak of streamflow (Cerdà, 1998; Neary et al., 2005; Shakesby and Doerr 2006), ultimately changing the catchments hydrological and sedimentary processes (García-Comendador et al., 2017; Ice et al., 2004; Zema, 2021). In addition, in light of current climate change, it is expected that extreme rainfall events may accelerate soil erosion in burnt areas.

Wildfires seriously increase soil erosion (Fernández and Vega, 2018; Viera et al., 2015) and impair surface
water quality by delivering fire-related contaminants to rivers with (Nunes et al., 2017; Verkaik et al., 2013;
Campos et al., 2012; Chessman, 1986; Olivella et al., 2006). Fire severity (amount and duration of

subsurface heating), nature of vegetation cover, physical and chemical characteristics of burnt areas (i.e.
climate, soil, topography), and the time interval between the fire and rainfall determine the degree of impact
on soil erosion and water quality (Viera et al., 2015; Tecle and Neary, 2015). Post-rehabilitation measures
are needed to mitigate the effects of fire on hydro-sedimentary response and protect soil from erosion
(Lucas-Borja, 2021). An accurate prediction of post-fire runoff and sediment yield is required to guide postfire risk management and plan soil and water restoration measures (Argentiero et al., 2021; Fernández et al.,
2010).

84 Hydrological and soil erosion models can provide valid support (Kampf et al., 2020) for quantifying the 85 catchment hydro-sedimentary response to forest fire events and planning adequate restoration measures. 86 Several modelling applications conducted to support management agencies are reported in the literature 87 (Lopes et al., 2021). The Revised Universal Soil Loss Equation (RUSLE) was applied to sites affected by fire in the Mediterranean Region to estimate the impact on runoff and soil erosion (Coschignano et al., 2019; 88 89 Efthimiou et al., 2020; Fernández et al., 2010; Lanorte et al., 2019; Rulli et al., 2013). Analogously, the Pan-European Soil Erosion Risk Assessment model (PESERA, Kirkby et al., 2004) was applied in central 90 91 Portugal (Esteves et al., 2012), Spain (Fernández and Vega, 2016), and Greece (Karamesouti et al., 2016). 92 The Water Erosion Prediction Project model (WEPP; Flanagan and Nearing, 1995) was used in Spain 93 (Fernández and Vega, 2018). Rulli and Rossi (2005; 2007) developed a distributed hydro-geomorphological 94 model to estimate the dynamics of fire-disturbed conditions at the basin scale. Di Piazza et al. (2007) used 95 the RUSLE model and a spatial disaggregation criterion for sediment delivery processes (SEDD model) to 96 assess the effects of bushfires in Italy. The Soil and Water Assessment Tool model (SWAT, Arnold et al., 97 1998) was applied in two Portuguese sites to estimate the post-fire impacts on streamflow and sediment yield 98 (Basso et al., 2020; Nunes et al., 2018). Grangeon et al. (2021) proposed the WaterSed model to simulate 99 forest fire and firebreak scenarios and analyse their respective effects on sediment loads. Zema et al. (2020) 100 adapted the Morgan-Morgan-Finney model after wildfires in Spain. However, Lopes et al. (2021) pointed out 101 in their review that many of the published studies reported modelling applications without field validation and that mitigation measures were simulated in a limited number of cases. The authors concluded that further 102 studies and tests were needed for adapting models to burnt conditions. Indeed, the model parametrisation is 103

not specifically designed for post-fire conditions, it needs to be adapted to the post-fire conditions, and
currently, it remains an open problem (Basso et al., 2020; Nunes et al., 2018).

In this context, the present paper contributes to bridging the gap in modelling post-fire impact and quantifying mitigation measures' effects. The general aim of the work is to test and adapt the SWAT model as a tool for rapid post-fire erosion risk assessment. The specific aims of this work are: (i) to simulate runoff and sediment yield for the current land use in a mountainous river basin, (ii) to predict the effects of forest fire on runoff, erosion, and sediment transport, and iii) to quantify the effects of post-fire mitigation measures on runoff and sediment yield at the basin, subbasin, and hydrologic response unit (HRU) scale.

Field measurements were used to calibrate the hydro-sedimentary parameters of the model for the current land use in the Celone (S-E Italy) river basin that is characterised by an intermittent river network feeding the Capaccio reservoir. The post-fire scenarios were simulated by changing the appropriate parameters affecting hydrological processes and soil erosion by water. The work provides useful post-fire management information (i.e. quantification) to the river basin managers.

# 117 **2.** Materials and methods

### **118 2.1 Study area**

The Celone River basin is located in northern Apulia (SE, Italy). The study area (72 km<sup>2</sup>) is located upstream
of the Capaccio reservoir (41° 25' 35''N; 15° 24' 52''E) (25.82 Million of m<sup>3</sup>), of which the Celone river is
the main inflow.

The elevation of the study area ranges from 1142 m a.s.l. to 218 m a.s.l. (mean value 386 m a.s.l.). Steep slopes characterise the upper part of the basin, making it prone to erosion. The main channel is incised in the mountainous area. Consequently, many check dams have been built to reduce bank erosion. Most of the coarse material is deposited in the first alluvial plain, resulting in a braided river. Downstream, it continues with a meandering pattern.

127 The lithology consists of flyschoidal units (flysch della Daunia), grey-blue clays in the mountain, and 128 alluvial deposits in the valley. The soils show a variable texture (clay, clay-loam, and sandy-clay-loam) and 129 are classified as typic-haploxerroll, vertic-haploxeroll, and typic-calcixeroll, according to the US Department



130 of Agriculture classification.

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Figure 1. Study area: Celone river basin (Apulia Region, Italy). a) DEM and subbasins distribution, subbasins affected by forest fire are delineated with continuous black lines (55, 63); b) Land use, gauging stations: MP (discharge and suspended sediment during the period 2010-2011), SV (discharge during the period 1994-1996).

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Mean annual rainfall is 770 mm (1990-2011), and mean temperature varies between 3.4°C (January) and
20.3°C (August) in the mountain, and between 7.2°C (January) and 25.5°C (August) in the valley (De
Girolamo et al., 2017a).

The soil erosion by water in the basin is both distributed (sheet erosion) and localised (rill erosion) (De Girolamo et al., 2015). It is favoured by agricultural practices such as conventional tillage (multiple operations with chisel plough and disks). The prevalent land use is for cereal growth (mostly winter and durum wheat; 45% of the catchment area). Other land use includes sunflowers (9%), natural degraded areas (6%), olive groves (8%), vineyards and vegetables (2%), and urban areas (1%). Forests, primarily oaks and conifers (29%), cover the mountainous part of the basin. The study area was monitored in 2010-2011 at the Celone Masseria Pirro gauge (41°23'41''N; 15°20'02''E) (MP in Figure 1), with continuous measurements of streamflow (Q) and discrete suspended sediment concentration (SSC) samples (De Girolamo et al., 2015; De Girolamo et al., 2018). Daily streamflow was computed starting from measurements taken on 15-min of the time step, and suspended sediment load at the monthly time scale was estimated using the sediment rating curve developed based on measured streamflow and suspended sediment concentrations (Eq. R3 in De Girolamo et al., 2018).

### 151 **2.2 Conceptual model**

The SWAT model with ArcGIS interface (Arnold et al., 1998) was used in the present work to simulate 152 153 streamflow and sediment yield and predict the potential impact of forest fire and post-fire measures on 154 sediment and hydrology. SWAT is a semi-distributed model able to predict hydrological processes, water 155 quality, and the environmental impact of land use and management practices on water bodies and soils in 156 agricultural basins (D'Ambrosio et al., 2020a; De Girolamo and Lo Porto, 2020). The SWAT model is 157 widely used for assessing the effects of anthropogenic pressures on water quality (Cakir et al., 2020; D'Ambrosio et al., 2020b; Pulighe et al., 2019) for estimating climate change impact on water resources and 158 159 flow regimes (Brouziyne et al., 2020; Glavan et al., 2015; Perra et al., 2018), and for simulating soil erosion 160 (Gamvroudis et al., 2015; Vigiak et al., 2017) and the impact of best management practices (BMPs) on water 161 resources (Ricci et al., 2020).

162 In SWAT, the basin is divided into subbasins that are further subdivided into HRUs, which are characterised 163 by homogeneous land use, soil, and slope. The water cycle is divided into the land phase and routing phase. 164 The components of the land phase (i.e. runoff, evapotranspiration, crop growth, soil erosion, nutrient and 165 pesticides loads entering into the main channel) and the methods used for their computation are described in 166 Neitsch et al. (2011). The routing phase through the river network includes transmission losses and 167 degradation of nutrients, pesticides, and bacteria. Similarly, the sediment budget is divided into two 168 components, landscape phase and channel routing. The soil erosion phase includes the detachment, transport, 169 and deposition of soil particles by the erosive force of raindrops and the surface flow of water. The channel 170 sediment routing phase considers deposition and degradation that occurs in the channel. The landscape 171 sediment phase is computed with the modified universal soil loss equation (MUSLE), and the channel

- sediment routing is computed using the Bagnold equation (Neitsch et al., 2011). The SWAT model provided
- 173 outputs at the basin, subbasin, and reach scale.
- The model was run at a daily time step from 1990 to 2011. The Hargreaves-Samani equation was selected for estimating the potential evapotranspiration (PET), and the SCS Curve Number Method was adopted to calculate surface runoff (Neitsch et al., 2011). Table I summarises input data used in the present study, their
- source, and resolution.
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179	Table I.	Input data:	variable.	source.	scale.	information.
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Variable	Source	Scale	Information
Precipitation	Civil Protection Service	Daily	2 weather stations
	Apulia Reg. Agency		(1990-2009)
Temperature	Civil Protection Service	Daily min, daily max	2 weather stations
	Apulia Reg. Agency		(1990-2009)
Land use map	Corine Land Cover 2000 EU Project	ArcInfo format (scale 1:100000)	Minimum area digitalized 25 ha
Soil map	ACLA 2 - FEOGA EU Project	ArcInfo format (scale 1:100000)	5 soil profiles
Management Practices	Consorzio per la Bonifica della	Subbasin scale; Municiality	Tillage oper., irrigation amount,
	Capitanata; farmers		fertilizers (type, timing, amount)
Digital Elev. Model	Apulia River Basin Authority	Arc Info grid format (8x8m)	
Streamflow (SV gauge)	Civil Protection Service	Daily	1992-1996
Streamflow (MP gauge)	Water Research Institute-National	Sub-daily	2010-2011
	Research Council (IRSA-CNR)		
Suspended sediment	Water Research Institute-National	Monthly	2010-2011
load (MP gauge)	Research Council (IRSA-CNR)		

# 181 **2.3 Model calibration**

182	The analysis of sensitivity, reported in De Girolamo et al. (2017a), identified among the most sensitive
183	parameters influencing hydrological processes the initial SCS curve number for moisture condition II (CN2),
184	the threshold depth of water in the shallow aquifer required for return flow to occur (GWQMN [mm]), the
185	available water capacity of the soil layer (SOL_AWC [mmH <sub>2</sub> O/mmsoil]), the soil evaporation compensation
186	factor (ESCO), the surface runoff lag time (SURLAG [days]), revap coefficient (GWREVAP), the Baseflow
187	alpha-factor (ALPHA_BF, [days]), and Groundwater delay time (GW_DELAY, [days]).
188	In the present study, the model SWAT2012 version was used. The basin was divided into 74 subbasins,
189	further partitioned into 200 HRUs. Conservation practices were not adopted in the study area (Panagos et al.,
190	2015a; Wischmeier and Smith, 1978). The conservation practice factor (USLE_P) was assumed to be equal
191	to 1 for all land uses, except for forested areas where the P factor was set to 0.8. According to the crop
192	systems, the crop management factor (USLE_C) was set within 0.0019 to 0.2, as suggested by Panagos et al.

193 (2015b).

194	Daily streamflow data were available at two gauging stations (SV and MP in Figure 1) over different time
195	periods (Table I). The model was calibrated for the daily streamflow at the SV gauge over 1994-1996 and at
196	the MP gauge over 2010-2011 (Figure 1). The sediment load (sediment transported by the river; SSC×Q)
197	was calibrated at the MP gauge at the monthly time scale (2010-2011), and the validation was carried out for
198	daily streamflow at the SV gauge (1992). Manual calibration was performed, including the above-mentioned
199	parameters for hydrology. For sediment load calibration, the following parameters were included: channel
200	erodibility factor (CH_COV1), channel cover factor (CH_COV2), Manning's "n" value for the main channel
201	(CH_N2), the maximum amount of sediment that can be transported from a river reach (SPCON), and the
202	exponent for calculating sediment that can be transported in the channel (SPEXP). Table II shows the
203	parameter values corresponding to the best fit for the most sensitive parameters and their range of variability.
204	The model's performance was evaluated by using the coefficient of determination ( $\mathbb{R}^2$ ), the Nash-Sutcliffe
205	efficiency (NSE), and the observation standard deviation ratio (RSR). The simulations were considered good
206	if $0.65 < NSE < 0.75$ , $0.5 < RSR < 0.6$ and $R^2 > 0.8$ and satisfactory if $0.5 < NSE < 0.65$ , $0.65 < RSR < 0.75$ , $0.75 < RSR < 0$
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207 and  $R^2 > 0.5$  (Moriasi et al., 2007).

200	Table II. Call	Table 11. Campiateu parameters (actual value useu) and then range of variability.							
	Parameter	Description	Actual value used	Range					
	CN2	SCS Curve number for moisture condition II	70-85 <sup>a</sup>	35-98					
	GWQMN	Threshold depth of water in the shallow aquifer required	800	0-5000					
		for return flow to occur [mm H <sub>2</sub> O]							
	OV_N	Manning's "n" value for overland flow	0.1-0.4 <sup>a</sup>						
	SOL_AWC	Available water capacity [mm H <sub>2</sub> O/mm soil]	0.12-0.21 <sup>a</sup>	0-1					
	ESCO	Soil Evaporation compensation factor	0.95	0-1					
	SURLAG	Surface runoff lag coefficient [days]	2	0-10					
	GWREVAP	Revap coefficient	0.02	0.02-0.2					
	ALPHA_BF	Baseflow alpha factor [days]	0.95	0-1					
	GW_DELAY	Groundwater delay time [days]	3						
	CH_N2	Manning's "n" value for main channel	0.11	0.05-0.5					
	CH_COV1	Channel erodibility factor	$0-0.5^{b}$	0-1					
	CH_COV2	Channel cover factor	0-5 <sup>b</sup>	0-1					
	SPCON	Maximum amount of sediment retrained during channel	0.007	0.0001-0.01					
		sediment routing							
	SPEXP	Exponent for calculating sediment retrained in channel	1.8	1-2					
209	<sup>a</sup> value varies according to input data (soil, land use)								

208 Table II. Calibrated parameters (actual value used) and their range of variability

210 <sup>b</sup> value = 0 was assumed for reaches in plain area; values > 0 was assumed in the mountainous and hilly reaches.

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# 212 **2.4 Analysis at the reach scale**

After the fire events, the sediment-associated pollutants transported via surface runoff could accumulate on the riverbed with several implications on water quality and ecological status. In order to identify the river segments where the deposition of sediment occurs, an analysis at the reach scale was carried out for the period 1990-2011. Thus, the sediment transported with water into the reach (SED\_IN) were combined with the sediment transported with water out of the reach (SED\_OUT), and the sediment from the subbasin to the river reach during the time step to identify the river reach's under erosion and deposition.

### 219 **2.5 Simulating post-fire scenarios**

Post-fire scenarios were simulated assuming that fire burnt the forest areas in two selected subbasins [Figure 1, Numbers 55, 63] since fire events not really occurred in the study. The basins were selected to analyse the effect of fire severity and management in soil erosion-prone areas. Both subbasins were characterised by steep slopes, high rainfall, and soil erodibility. The drainage area of subbasin 55 was 2.09 km<sup>2</sup>, 1.17 km<sup>2</sup> was covered by forests, and the drainage area of subbasin 63 was 2.47 km<sup>2</sup>, including 0.47 km<sup>2</sup> covered by forests. The fire was modelled to occur in 2010 only in forested areas.

226 The post-fire scenarios and the effectiveness of selected mitigation measures in reducing soil erosion by 227 water were simulated and results were analysed for one year after the fire (data on the post-fire measures did not exist in the basin). Mayor et al. (2007) and Pausas et al. (2008) pointed out that soil erosion might be two 228 orders of magnitude higher five years after the fire. However, the highest hydrological and erosive events 229 230 occur beyond the first year after the fire (García-Comendador et al., 2017). The following six scenarios were simulated to provide a wide range of potential impacts on hydro-sedimentary response to support post-fire 231 232 management. The model parameters influencing runoff and soil erosion were properly modified for each 233 scenario using literature values. Table III shows the parameters and their values for the baseline and post-fire 234 scenarios and the most relevant references used as guides for their selection and values' attribution.

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#### 236

### Scenario Fr1: high-severity fire and post-fire logging

It was assumed that "high-severity fire" was ground and canopy fire (all shrubs and herbaceous plants killed)
with high soil heating and alteration of soil structure (decreased infiltration and increased water repellency).
This scenario analysed the potential effect of removing fire-killed trees from burnt areas (logging) and the
successive tillage operation (chisel plough) on those areas.

The fire effect on soil characteristics was simulated by modifying the USLE erodibility factor (USLE\_K).
The effect of fire on soil water repellency (Sol\_K) was incorporated into the USLE\_K by adopting the

suggestions by Miller et al. (2003) (reported by Larsen and MacDonald, 2007). The reduction of soil
protection due to the damage of vegetation cover was considered by modifying USLE\_C. In literature, the
post-fire USLE\_C factor applied ranges from 0.01 (low severity) to 0.3 (high severity) (Borrelli et al., 2016).
USLE\_P was set equal to one, and the increase of runoff at the different spatial scales consequent to the fire
events were estimated by modifying OV\_N and the CN2 (Table III).

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- 249

#### Scenario Fr2: high-severity fire and natural regeneration

High-severity fire impact on soil was simulated by increasing the USLE\_K (Table III). For this scenario,
USLE\_P was set to 1, and USLE\_C was fixed to 0.13 to mimic the effect of the regrowth of vegetation. CN2
was increased (+15) compared to the baseline scenario (Havel et al., 2018). Meanwhile, OV\_N was assumed
lower than the baseline (Table III).

254

#### 255 *Scenario Fr3: high-severity fire and emergency stabilisation (straw mulching and seeding)*

Straw mulching was considered in this scenario to protect soil after the fire. The effect of straw mulching was simulated by modifying USLE\_P, USLE\_C, CN2, and OV\_N. USLE\_P was set to 0.343 and USLE\_C to 0.13, considering the effect of seeding and regrowth of vegetation (Fernandez et al., 2010; and Rulli et al., 2013). In addition, mulch material on soil is a conservation practice that is generally simulated by modifying the CN2; here, it was reduced by 3 points compared to the value assigned in Fr2 (Waidler et al., 2009). Finally, OV\_N was increased compared to Fr1 and Fr2, but lower than the one assumed in the baseline, as suggested by Neitsch et al., 2011.

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#### 264

#### Scenario Fr4: moderate-severity fire and erosion barriers

In this scenario, the moderate-severity fire was hypothesised, and erosion barriers were simulated as a postfire mitigation measure to reduce surface runoff and soil losses. The assumption was ground fire and burning of lower tree limbs, moderate soil heating, increased water repellency and decreased infiltration. The baseline value of USLE\_K was assumed to increase (Table III), and OV\_N and CN2 were reduced and increased, respectively. USLE\_P was modified by adopting the value of 0.85 (Myronidis et al., 2010). 270 Nevertheless, it is important to underline that literature reports a wide range of USLE\_P values applied (from271 0.2 to 0.85).

272

### 273 Scenario Fr5: low-severity fire and natural regeneration

Low-severity fire and natural regeneration were simulated in this scenario. It was assumed that leaf litter was completely consumed with small changes in soil properties. All the parameters mentioned above were modified as reported in Table III. OV\_N was assumed to be 0.3, and the baseline value of CN2 was increased (+5), USLE\_K was assumed to increase to a lesser extent than moderate-and high-severity fire.

278

### 279 Scenario Fr6: very low-severity fire and natural regeneration

In this scenario, it was assumed that fire had very lightly charred only fine fuel and litter on the ground and soil properties (i.e. hydraulic saturated conductivity, water repellency) were unchanged. The baseline value of CN2 was slightly increased (+3), and USLE K was unchanged (Table III).

#### 283 Table III. SWAT parameters used in the baseline simulation and fire scenarios.

Parameter	Description	Baselin e	Fr1 High- severity fire, logging, tillage	Fr2 High- severity fire, natural regeneration	Fr3 High severity fire, straw mulching and seeding	Fr4 Moderate- severity fire and erosion barriers	Fr5 Low- severity fire and natural regeneratio n	Fr6 Very low- severity fire and natural regeneratio n	Reference
OV_N	Manning's roughness coefficient	(0.4)	0.09 <sup>a</sup>	0.16 <sup>a,b</sup>	0.22 <sup>a</sup>	0.25 <sup>a</sup>	0.3ª	0.3ª	<sup>a</sup> Neitsch et al., 2011; <sup>b</sup> Stoof et al. 2015
USLE_P	USLE eq. supporting practice factor	0.8 <sup>a,b</sup>	1	1	0.343 <sup>c.d</sup>	0.85 <sup>d,e</sup>	0.9	0.9	<sup>an</sup> Panagos et al., 2015a; Wischmeier and Smith, 1978; <sup>c</sup> Fernandez et al., 2010; <sup>d</sup> Rulli et al., 2013; <sup>e</sup> Myronidis et al., 2010
USLE_C (m <sup>2</sup> /m <sup>2</sup> )	USLE C factor for water erosion	0.0019 <sup>a</sup>	0.23 <sup>b</sup>	0.13 <sup>c.d</sup>	0.13 <sup>c.d</sup>	0.05 <sup>e.f</sup>	0.01 <sup>e,f</sup>	0.01 <sup>e,f</sup>	<sup>a</sup> Panagos et al., 2015b; <sup>b</sup> Fernandez et al., 2016; <sup>c</sup> Fernández & Vega, 2018; <sup>d</sup> Rulli et al., 2015; <sup>e</sup> Larsen and MacDonald, 2007; <sup>f</sup> Terranova et al. 2009
CN2	Initial SCS runoff curve number for soil moisture condition II	(70)	90 <sup>a</sup>	85 <sup>b</sup>	82 <sup>b.c</sup>	80 <sup>b,d</sup>	75 <sup>b.d</sup>	73	<sup>a</sup> Neitsch et al., 2002; <sup>b</sup> Havel et al., 2018; <sup>c</sup> Waidler et al., 2009; <sup>d</sup> Basso et al., 2020.

USLE_K* (ton acre hour)/(hun dred acre ft ton inch)	USLE eq. Soil erodibility factor	0.13- 0.15	+0.016/0.131 7 <sup>a.c.d</sup>	+0.016/0.131 7 <sup>a</sup>	+0.016/0. 1317ª	+0.015/0.1 317 <sup>b</sup>	+0.014/0.1 317 <sup>c</sup>	0.13-0.15 <sup>e</sup>	<sup>a</sup> Miller et al., 2003; <sup>b</sup> Fernandez et al., 2010; <sup>c</sup> Basso et al., 2019; <sup>d</sup> Coschignano et al., 2019; <sup>c</sup> Di Piazza et al., 2007.
Tillage	Plowing (chisel plow)		Deep						Nunes et al., 2018.

\*0.1317 is the conversion factor for soil erodibility factor (USLE\_K) from t h MJ<sup>-1</sup> mm<sup>-1</sup> to ton acre hour/ hundred acre ft ton inch

### 285 **3. RESULTS**

### 286 **3.1 Modelling streamflow and sediment load**

The statistics obtained for daily streamflow calibration showed a good model efficiency at the SV gauge 287 (NSE = 0.70; RSR = 0.54;  $R^2 = 0.88$ ) and at the MP gauge (2010-2011) (NSE = 0.73; RSR = 0.50;  $R^2 =$ 288 0.89). Similar results were obtained for the validation period of the streamflow at the SV gauge (NSE = 0.73; 289 RSR = 0.63;  $R^2 = 0.90$ ). Figure 2 shows the simulated and observed streamflow for the calibration period at 290 the SV gauge. The performance for sediment calibration on the monthly time scale at the MP gauge was 291 satisfactory (NSE = 0.73; RSR = 0.70;  $R^2 = 0.54$ ). The results showed an underestimation of sediment load 292 293 in March 2011, when a series of large floods occurred, and an overestimation in autumn. Figure 3 shows the observed and simulated daily streamflow at the MP gauge (Figure 3A) and monthly observed specific 294 sediment load (SSL, t ha<sup>-1</sup>) versus simulated values (Figure 3B). 295





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302 Figure 3. Observed and simulated daily streamflow at the MP gauge (A). Measured versus simulated specific sediment load SSL (t ha<sup>-1</sup>) at monthly time scale for the calibration period (2010-2011) (B).

#### **3.2 Streamflow and sediment load for the current land use (baseline)** 305

At the basin scale, from 1990 to 2011, the average annual rainfall was 777 mm (SD = 179 mm), mainly 306 307 concentrated from November to April (wet season), the surface runoff was 114 mm (SD = 66 mm), and the total water yield was 288 mm (SD = 140 mm). Most of the rainfall (61%) was lost via actual 308 evapotranspiration (471 mm; SD = 41 mm), and the potential evapotranspiration was 954 mm (SD = 30309 mm). The average annual SSY (sediment yield per unit of catchment area and unit of time; t ha<sup>-1</sup>yr<sup>-1</sup>) was 310 5.60 t ha<sup>-1</sup> yr<sup>-1</sup> (SD = 3.47 t ha<sup>-1</sup> yr<sup>-1</sup>). A high inter-annual variability characterised all the water balance 311 312 components and the SSY due to differences in climate conditions. In the driest year (2000), the total annual rainfall was 471 mm, surface runoff (SR) was about 26 mm, and the SSY was 3.03 t ha<sup>1</sup> yr<sup>-1</sup>. In the wettest 313 year (2009), the total annual rainfall was 1217 mm, SR was 300 mm, and SSY was 13.82 t ha<sup>-1</sup> yr<sup>-1</sup>. 314



#### Figure 4. Average specific sediment yield (SSY, t ha<sup>-1</sup> yr<sup>-1</sup>) at the subbasin scale simulated from 1990 to 2011.

At the subbasin scale (Figure 4), over the period 1990-2011, the mean annual SSY was < 1.4 t ha<sup>-1</sup> yr<sup>-1</sup> in the 318 subbasins located in the plain area (14% of total drainage area). Most of the subbasins showed values 319 between 1.4 to 10 t ha<sup>-1</sup> yr<sup>-1</sup>, and some mountainous subbasins (20% of total drainage area)—characterised by 320 steep slopes—showed severe soil erosion (SSY > 10 t ha<sup>-1</sup> yr<sup>-1</sup>). These results are consistent with the new 321 assessment of soil loss by water erosion in Europe performed with the RUSLE2015 by Panagos et al., 2015c 322 who reported soil loss rate in the range of 0 - 2 t ha<sup>-1</sup> yr<sup>-1</sup> in lowland, values from 2 t ha<sup>-1</sup> yr<sup>-1</sup> to 5 t ha<sup>-1</sup> yr<sup>-1</sup> in 323 hilly areas, and values mostly in the range of 5 t ha<sup>-1</sup> yr<sup>-1</sup> to 10 t ha<sup>-1</sup> yr<sup>-1</sup> with several cells showing soil loss 324 rate > 10 t ha<sup>-1</sup> yr<sup>-1</sup> in the mountainous part of the basin. 325

At the HRU level, natural degraded areas, predominant in the steep slopes areas, showed the highest values of SSY. Also, durum wheat fields, where up-and-down tillage was generally adopted, showed high values of SSY. Garrigue, deciduous, and mixed forests showed lower values of SSY. The box plot in Figure 5 shows the annual SSY estimated at the HRU level for each crop. Wide variability was found among the HRUs for each crop because of the different environmental factors (slope, soil, rainfall) that influence hydrology and soil erosion.

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Figure 5. Box plot of the average specific sediment yield (SSY, tha<sup>-1</sup>yr<sup>-1</sup>) simulated at the HRU level from 1990 to 2011. The horizontal line within the box plot indicates the median, boundaries of the box indicate the 25<sup>th</sup> and 75<sup>th</sup> percentile, and whiskers indicate the minimum and maximum values.

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The reach-scale analysis for identifying the river segments where sediment deposition occurs showed that most first-order river segments were under erosion. Meanwhile, sediment deposition was predicted in some intermediate reaches and those located in the alluvial plains (Figure 6). In the latter, if fire events occur in the upstream areas, pollutants such as Fe, Mn, As, Cr, Al, Ba, and Pb could be deposited along the river bed, and the water quality could be impaired (Smith et al., 2011).



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348 Figure 6. Segments of the Celone River under erosion and deposition.

## 349 3.3 Post-fire scenarios: potential impact on hydro-sedimentary response

At the basin scale, the integrated effect of the two burnt areas (1.64  $\text{km}^2$ , 2.3% of the entire river basin) on 350 351 surface runoff was negligible. Only the scenario Fr1 showed a slight increase in annual surface runoff (99.95 352 mm) compared to the baseline (99.53 mm). Similarly, the impact of wildfire on the total water yield (total streamflow at the outlet for the unit area; TWY, mm) was negligible for all the scenarios. The lateral flow 353 and baseflow contributions to the streamflow showed a slight decrease only for Fr1 (194.56 mm) compared 354 to the baseline (194.88 mm). It can be inferred that these results depend on the limited extension of the burnt 355 356 area (2.3%). For all the fire scenarios, including post-fire mitigation measures, an increase in SSY was modelled, ranging from 5.86 t ha<sup>-1</sup> yr<sup>-1</sup> (baseline) to 12.05 t ha<sup>-1</sup> yr<sup>-1</sup> (Fr1) (Figure 7A). The severity of the fire 357 played an essential role in SSY. A massive difference was predicted between high-severity fire (Fr1 and Fr2) 358 and low-severity fire scenarios (Fr5, Fr6, Figure 7A). Fr5 and Fr6 showed limited increases in SSY (6.4 and 359 6.3 t ha<sup>-1</sup> yr<sup>-1</sup>) compared to the baseline. The post-fire management decreased SSY compared to Fr1 (8.9 t ha<sup>-1</sup> 360



Figure 7. Specific sediment yield (SSY, t ha<sup>-1</sup> yr<sup>-1</sup>) simulated for the baseline and post-fire scenarios at the basin scale (A), subbasin scale (B), and at the hydrological response unit level (HRU) (C).

Results at the subbasin scale showed negligible variations in surface runoff (ranging from 129.01 mm to 371 372 129.17 mm) for all the analysed scenarios in the subbasin 55 compared with the baseline (129.00 mm). 373 Similarly, the increase in surface runoff simulated for the subbasin 63 was negligible, ranging from 98.78 mm (baseline) to 98.83 mm (Fr1), and for low-severity fire simulated in the Fr5 and Fr6 scenarios, it was 374 98.79 mm. The SSY simulated for the baseline (9.5 t ha<sup>-1</sup> yr<sup>-1</sup> and 9.7 t ha<sup>-1</sup> yr<sup>-1</sup>, for sub 55 and sub 63, 375 respectively) increased up to 57.4 t ha<sup>-1</sup> yr<sup>-1</sup> (sub 55, Fr1) and up to 26.1 t ha<sup>-1</sup> yr<sup>-1</sup> (sub 63, Fr1), confirming 376 that the high severity of fire events and the post-fire logging may produce a dramatic increase in soil loss 377 (Figure 7B). The extension of the burnt area within the basin played an essential role in SSY variations. 378 Indeed, as a result of the larger burnt area in the subbasin 55 (56%), the SSY predicted in this subbasin was 379 380 much more than SSY simulated in the subbasin 63 (burnt area 19% of subbasin area), especially in the high-381 severity fire scenarios with no measures to reduce soil erosion (Fr1, Fr2) (Figure 7B). Post-fire mitigation treatments (Fr3 and Fr4) effectively reduced soil erosion in high- and moderate-severity fires. In particular, 382 383 straw mulching and seeding—as simulated in Fr3—protected ground cover better than erosion barriers (Fr4) (Figure 7B). Indeed, SSY was19.11 t ha<sup>-1</sup> yr<sup>-1</sup> and 13.13 t ha<sup>-1</sup> yr<sup>-1</sup> (subbasin 55 and 63, respectively) in Fr3 384 and 20.93 t ha<sup>-1</sup> yr<sup>-1</sup> and 14.35 t ha<sup>-1</sup> yr<sup>-1</sup> (sub 55 and 63, respectively) in Fr4, while fire severity was 385 386 simulated to be high and moderate, respectively. As expected, due to the lower severity of fire represented by

the Fr5 and Fr6, SSY increased to a lesser extent in these scenarios, ranging from 11.4 t ha<sup>-1</sup> yr<sup>-1</sup> (Fr6) to 13.3 387 t ha<sup>-1</sup> yr<sup>-1</sup> (Fr5) in the subbasin 55, and from 10.62 t ha<sup>-1</sup> yr<sup>-1</sup> (Fr6) to 11.57 t ha<sup>-1</sup> yr<sup>-1</sup> (Fr5) in the subbasin 63. 388 389 The analysis of the potential impact of post-fire scenarios in terms of soil erosion was carried out also at the HRU level. Figure 7C shows the results for the three HRUs. The SSY estimated for the baseline ranged from 390 1.18 t ha<sup>-1</sup> yr<sup>-1</sup> to 2.04 t ha<sup>-1</sup> yr<sup>-1</sup>. It increased more than one order of magnitude for the high-severity fire 391 scenarios, Fr1 ranged from 78.19 t ha<sup>-1</sup> yr<sup>-1</sup> to 95.77 t ha<sup>-1</sup> yr<sup>-1</sup> and Fr2 from 49.40 t ha<sup>-1</sup> yr<sup>-1</sup> to 59.91 t ha<sup>-1</sup> yr<sup>-1</sup> 392 <sup>1</sup>. As expected, the very low-severity fire scenario presented the lower increase of SSY, ranging from 4.33 393 (HRU 2,55) to 6.74 t ha<sup>-1</sup> yr<sup>-1</sup> (HRU 3,63) (Figure 7C). 394

### 395 **4. Discussion**

### **396 4.1 Simulating baseline**

397 Soil erosion models are widely used around the world for estimating soil losses by water (Borrelli et al., 2021, Bezac et al., 2021), although some critical points have not been completely solved yet (i.e. 398 399 parameterisation, lack of measurements to validate results, upscaling from local to larger scales). The implementation of a conceptual model is a complex activity that requires a broad knowledge of the river 400 basin and the integration of various data and knowledge sources. In addition, conceptual models must be 401 402 applied by experienced modellers who have deep knowledge of hydrological and sedimentary processes (i.e. equations, parameters). The present study shows that the SWAT model is a valuable tool to simulate both the 403 hydrological processes and SSY under the Mediterranean climate, and it has great potential in watershed 404 management. 405

406 The model has already been successfully used in the Apulia Region (De Girolamo et al., 2017a,b) for 407 analysing hydrological processes. However, the low flow was generally overestimated. Similarly, in the present work, the model did not simulate the zero-flow condition, which was recorded in the summer in 408 several observed years. The minimum flow predicted by the model was about 20 1 s<sup>-1</sup>. In the previously 409 mentioned studies, which were oriented to support ecological status evaluation, it was identified that a zero-410 flow threshold and time series of streamflow were appropriately modified. In the present study, taking into 411 account that the extremely low flow is characterised by negligible sediment transport, the discrepancy 412 between observed and simulated streamflow was considered insignificant for the research. 413

The model performance in simulating SSY was satisfactory. Nevertheless, SSY was underestimated in the extremely wet conditions and slightly overestimated in autumn, confirming the results obtained by Abdelwahab et al. (2018), who implemented the SWAT model in the Carapelle basin (Apulia Region). Data resolution and problems linked to the transferability of the Modified Universal Soil Loss Equation approach may have influenced model performances (Ricci et al., 2018; Williams & Berndt, 1977).

At the basin scale, over the period 1990 to 2011 that included both dry and wet years, SSY was 5.60 t ha<sup>-1</sup> yr<sup>-</sup> 419 <sup>1</sup>. This estimate was comparable with the studies carried out in the same region by Ricci et al. (2018). At the 420 subbasin scale, SSY varied in the range 0.2-17.6 t ha<sup>-1</sup> yr<sup>-1</sup>. 20% of the total drainage area presented SSY 421 values higher than the critical value (10 t ha<sup>-1</sup> yr<sup>-1</sup>). These results agree with the soil losses estimated by water 422 erosion in Europe. Indeed, the application of a modified version of the RUSLE model (RUSLE2015) 423 424 (Panagos et al., 2015c; 2016) and the PESERA model (Kirkby et al., 2004; 2008) estimated soil losses rate in the range 0 - 2 t ha<sup>-1</sup> yr<sup>-1</sup> in lowland, values from 2 t ha<sup>-1</sup> yr<sup>-1</sup> to 5 t ha<sup>-1</sup> yr<sup>-1</sup> in hilly areas and values generally 425 from 5 t ha<sup>-1</sup> yr<sup>-1</sup> to 10 t ha<sup>-1</sup> yr<sup>-1</sup> in the mountainous part of the basin where several cells showed soil loss rate 426  $> 10 \text{ t ha}^{-1} \text{ yr}^{-1}$ . 427

428 At the HRU's level, land use and management practices played a key role in determining SSY variations. Natural degraded areas with a very low vegetation rate showed a very high SSY (median value 23.8 t ha<sup>-1</sup> yr<sup>-</sup> 429 <sup>1</sup>), mainly due to their location on steep slopes. Agricultural lands predominated by the basin's prevalent 430 crop—durum wheat—showed a median value of 3.2 t ha<sup>-1</sup> yr<sup>-1</sup>, comparable with the predicted soil loss rate 431 432 from erosion plots in Europe (Cerdan et al., 2010). Deciduous and mixed forests showed low SSY (0.3 t ha<sup>-1</sup>  $vr^{-1}$ ). These results were expected since it is well known that human activities such as agriculture and land-433 434 use change have induced an important increase in erosion rates (Foucher et al., 2021; Poesen, 2018). In the study area, soil losses are favoured by up and down ploughing, which is common, especially in mountainous 435 436 areas.

It is important to remember that the dataset used for sediment calibration was limited and that measurements taken at the outlet could be insufficient for optimal parameterisation. Hence, an uncertainty degree could affect the results at the subbasin and HRU levels. In the present study, parameters such as USLE\_P and USLE\_C were fixed on the literature basis and were not calibrated. A new monitoring plan with a nested approach could be very useful for improving model parameterisation and SSY estimation. 442 Despite its limitations, the model can predict the hydro-sedimentary response of the basin and may 443 contribute to the management of the reservoir, providing both the inflow and sediment loads.

### 444 4.2 Simulating post-fire mitigation measures

445 The forest located in the upper Celone river basin is an important natural area, it has been recognised as a Site of Community Importance, and it is included in the network "Natura 2000" (IT9110003) that covers 446 Europe's valuable species and habitats. The Regional Plan 2018-2020 developed by the Civil Protection 447 448 Agency classified this area as a "high risk of fire" site due to weather conditions and ignition sources (Civil 449 Protection Agency, 2018). Future climate projections predict an increase in temperature and a reduction of 450 rainfall (De Girolamo et al., 2017b) that could increase the probability of wildfires and the risk of short and 451 long-term post-fire contamination for surface waters. Hence, studies on the effects of wildfire on hydrology 452 and soil erosion (Zema, 2021) as well as the analysis of different post-fire scenarios (Rulli et al., 2013) are 453 needed to support watershed managers in managing the post-fire risks.

The hydro-sedimentary response of a watershed to fire events is complex (Vieira et al., 2018). It is related to fire impact on soil properties and changes in the vegetation cover (Cerdà and Doerr, 2008; Neary et al., 1999; Neary et al., 2005; Shakesby et al., 2011). The difficulties in evaluating the hydrological and sediment regimes generally increase in the Mediterranean environment with intermittent river networks due to the high spatial variability of soil properties, land use, and climate (Forteza et al., 2021).

The SWAT model, indispensable in water and soil management, may be used for the scenario analysis in the context of wildfire, but it needs to be adapted. Indeed, SWAT and all other hydrological/soil erosion models have not been developed specifically to simulate post-fire conditions. The adaptation consists of changing hydrological, soil, and cover parameters in an attempt to mimic the effect of fire (Lopes et al., 2021). Then, the model predictions should be calibrated, comparing the results with measurements. This is a critical point; most studies have not been validated with field observations since the latter are rarely available, especially at the basin scale (Lopes et al., 2021).

After fire events, land degradation and soil properties changes are not easy to measure and model since the
effect may change according to the severity of fire and characteristics of the soils (Neary et al., 1999).
Literature reports severe impacts on soil properties, providing sometimes conflicting results. Ice et al. (2004)

reported that reduction in infiltration rate could be very high (i.e. one or two orders of magnitude). Stoof et 469 470 al. (2015), in their study in Portugal, evaluated that despite the high fire intensity, bulk density, organic 471 matter, porosity, and saturated hydraulic conductivity did not significantly change. Nevertheless, they 472 concluded that even if the fire has a low impact on soil properties, it may have a high impact on runoff and erosion. Mataix-Solera et al. (2011), in their review, reported that the effect on soil aggregate stability may 473 increase or decrease for similar fire-severity events according to the soil characteristics. Post-fire water 474 475 repellency, which is a key factor in post-fire erosion since it reduces infiltration rate, especially after high-476 severity fires, is highly variable spatially (Doerr et al., 2009; Shakesby and Doerr, 2006) and difficult to accurately estimate. 477

478 This work assumes that wildfire increases the soil's water repellency and reduces the saturated hydraulic 479 conductivity, except for the very-low fire (Fr6). A reduction of the soil protection consequent to damage of vegetation cover was assumed to vary according to the scenarios. Hence, the effect of fire and the post-fire 480 mitigation measures on runoff and SSY was estimated by modifying parameters such as OV N, CN2, 481 USLE K, USLE C, and USLE P after an accurate literature analysis. To take into account the change in 482 483 water repellency (not explicitly considered in the models) and the consequent reduction of soil permeability, it was assumed an increase of USLE K by 0.016 Mg ha<sup>-1</sup> MJ<sup>-1</sup> mm<sup>-1</sup> ha h, considering a high rate of change 484 485 in soil erodibility (60-80%) as suggested by Miller's et al. (2003).

486 Post-fire measurements were not available. Due to this, the model was not calibrated for the above-487 mentioned scenarios, and the parameters were fixed based on the literature. This is a limitation of the present 488 study because the parameters adopted based on measurements made in other Mediterranean sites do not 489 necessarily apply to the Celone river basin. The choice of the scenarios was performed keeping in mind the 490 necessity of providing a wide range of realistic effects of wildfire and mitigation measures on soil loss and 491 runoff. Consequently, the above-mentioned parameters changed dramatically too. Thus, USLE\_K was 492 assumed to vary from +80% for high-severity fire to no difference for the very low-severity fire. They were 493 the highest and lowest values reported in the literature, respectively. Similarly, CN2 was assumed to change drastically (73 to 90 for Fr6 and Fr1, respectively). 494

495 According to the assumptions, wildfires have an important effect on the sedimentary response. The 496 increment related to the runoff was negligible in all the analysed scenarios. Lucas-Borja et al. (2019) 497 highlighted that the type of treatment (i.e. mulching or logging) did not influence the runoff generation in 498 their plots. Fr1 and Fr2 showed a dramatic increase in SSY for the three HRUs analysed, increasing in the 499 worst case (HRU 1, Sub. 55) from 1.26 t ha<sup>-1</sup> yr<sup>-1</sup> (baseline) to 95.8 t ha<sup>-1</sup> yr<sup>-1</sup> and to 59.9 t ha<sup>-1</sup> yr<sup>-1</sup>, 500 respectively (Figure 7). Malvar et al. (2017) and Wagenbrenner et al. (2015) evidenced that logging 501 operations may increase SSY mainly because of the trail generated by the passage of heavy machinery.

Fr5 and Fr6 showed a moderate increase of SSY that was quantified in 8.6 and 5 t ha<sup>-1</sup> vr<sup>-1</sup> (HRU 1, Sub. 55). 502 503 respectively. These results agree with the Shakesby (2011) studies, which pointed out that from high to low-504 severity fire, the effect on erosion may vary from more than two orders of magnitude or may not show differences at all. From the modelling point of view, the difference in SSY between Fr5 and Fr6 was mainly 505 attributable to the USLE K factor and, to a very small extent, to CN2 (-2 in Fr6) since all the other 506 507 parameters were unchanged. This result confirmed the USLE\_K factor as a very sensitive parameter in soil loss modelling. The difference in SSY between the Fr1 and Fr2 resulted from the integrated effect of several 508 509 parameters (USLE C, CN2, and OV-N) that were differentiated in the two scenarios (Table III).

The post-fire mitigation measures have been widely implemented, but the assessment of their efficiency has 510 511 been limited to local studies mainly conducted at the hillslope scale (Girona-García et al., 2021). The authors highlighted the need for studies on post-fire erosion mitigation measures, especially in high soil erosion 512 areas. In the present study, the mulching treatment (Fr3) reduced SSY (20.2 t ha<sup>-1</sup> yr<sup>-1</sup>) compared with the 513 514 high-severity fire Fr2 producing a reduction of SSY (66%). This result confirmed the study by Fernandez et 515 al. (2011) carried out in Galicia, where the authors concluded that straw mulch application with 80% soil 516 cover reduced soil loss by 66%. Fr3 resulted in more effective mitigation than the moderate-severity fire and erosion barriers. From the modelling point of view, this result is mainly attributable to the parameter 517 USLE P, which was assumed equal to 0.343 for straw mulching (Fernandez et al., 2010). When moderate-518 severity fire and erosion barriers were modelled (Fr4), SSY ranged from 19.8 to 26.2 t ha<sup>-1</sup> yr<sup>-1</sup> in the three 519 520 analysed HRUs showing a reduction (56-61%) compared with Fr3. These results agree with the study by Rulli et al. (2013), who determined a value of 24.1 t ha<sup>-1</sup> yr<sup>-1</sup>, and with Fernandez et al. (2011), who observed 521 a mean efficiency of barriers in retaining sediment of 58%. 522

### 523 **4.3 Future perspectives**

Despite the limits of the present study, the results clearly indicate that the rate of soil loss for the current land use and management practices is much higher than the soil rate formation that was estimated for European soils in 140 t km<sup>-2</sup> y<sup>-1</sup> (0.056 mm y<sup>-1</sup>) by Verheijen et al. (2009). This study confirms that it is urgent to reverse this trend by promoting soil loss mitigation measures (Montanarella and Panagos, 2021).

Ricci et al. (2020) analysed the efficiency and economic implications of some best management practices (BMPs) like contour farming, no-tillage, and reforestation, for the public and private sectors. They concluded that those BMPs, which the Apulia Region Rural Development Programme currently supports, effectively reduce soil losses but have not yet been adopted at a large scale. Several barriers still exist that limit their adoption (e.g., farmers' education, lack of awareness of soil erosion). Numerous actions are needed to favour the adoption of BMPs, and important public economic resources are needed to support a plan for soil protection.

In order to address these challenges, the EU's common agricultural policy may have an important role in ensuring that agriculture is in line with the soil protection principles. The new European Green Deal (EGD) with the "Farm to Fork" and the "zero pollution action plan" strategies will be central in preserving soil systems and biodiversity (Montanarella and Panagos, 2021). Research and monitoring may play an important role in reaching the EGD's goals.

540 In the next decades, increased fire risk is expected in the Mediterranean. Watershed management will need fire prevention efforts and specific actions to protect and restore the river basins before disturbance occurs. 541 542 95% of fires are due to human activities (i.e. agricultural practices) or negligent behaviour and arson (Vilar del Hoyo et al., 2009). It is, therefore, necessary to increase public perception and awareness of the risks of 543 wildfires and their impact on soil and water resources. Fire impact on soil is significant (Cerdà and 544 545 Robichaud, 2009), leading to an increase in soil erosion (Shakesby and Doerr, 2006). Hence, implementing 546 mitigation measures to reduce soil erosion is imperative and should be a part of every forest and soil recovery strategy (Bento-Gonçalves et al., 2012). This study has shown the effectiveness of straw mulching, 547 548 seeding, and soil erosion barriers in reducing soil erosion. However, further studies and new monitoring programs are needed to assess additional mitigation measures and adequately analyse their cost-549 550 effectiveness.

### 551 **5.** Conclusions

Through a case study, the present work contributes to bridging the gaps in modelling post-fire impact and mitigation measures on soil erosion and the hydrological response of a Mediterranean watershed with an intermittent river network.

555 The SWAT model, calibrated with field measurements, was applied for the current land use and land 556 management practices for hydrology and sediment yield, then it was applied to simulate post-fire mitigation 557 options. The model adequately reproduced the measured discharge for two monitoring periods: 1994-1996 558 and 2010-2011. It also satisfactorily reproduced suspended sediment dynamics over the period 2010-2011, 559 indicating that it may be used to analyse the hydro-sedimentary response of the basin. At the basin scale, 560 results showed that the average soil loss estimated over a long period (1990-2011) is much higher than the 561 soil formation rate. About 20% of the drainage area showed critical values of soil losses. At the subbasin 562 scale and HRU level, the results showed that steep slopes areas and natural degraded areas are under severe soil erosion. However, results could be affected by a large uncertainty since the dataset used for sediment 563 564 calibration was limited and the measurements of sediment taken only in a river section located in lowland 565 could be insufficient for optimal parameterization.

The results reveal the need of implementing mitigation measures to reduce soil losses. Water resources managers are called to take action to increase the awareness of farmers on soil erosion-related problems and to develop strategies to promote the adoption of BMPs such as contour farming, no-tillage, and reforestation. On the other hand, programs of soil conservation must be implemented over long time frames with important investments with strong actions of the policymakers at the European, national and regional levels.

Due to weather conditions and ignition sources, the basin is classified as a "high risk of fire" site. The probability of wildfires and the risk of short and long-term post-fire contamination of surface water could increase due to climate change in the near future. The results of the present study clearly showed that watershed management may have an important role in reducing the effects of wildfire on soil and water by implementing post-fire risk mitigation and restoration measures. SWAT—a hydrological and water quality model—may contribute to selecting the mitigation options to reduce soil erosion after a fire. In addition, the SWAT model is also a useful tool for the post-fire risk assessment in terms of water quality since it identifies the river segments where sediment-associated pollutants transported via surface runoff could accumulate onthe riverbed after fire events.

580 The choice of the post-fire scenarios was performed keeping in mind the necessity of providing a wide range of realistic effects of wildfire and mitigation measures on soil loss and runoff. According to the assumption, 581 high-severity fire vastly increases SSY at the basin and subbasin scales and HRU levels. This study shows 582 that a dramatic increase in soil erosion occurs in areas sensitive to erosion, demonstrating that major efforts 583 584 are needed to prevent forest fires and better manage the post-fire. The results showed that a small part (2%) 585 of the catchment is enough to cause a dramatic increase in soil loss quantified at the basin scale by up to 12 t ha<sup>-1</sup> vr<sup>-1</sup>. Post-fire management is effective at mitigating fire impact on soil erosion. In particular, post-fire 586 587 mitigation measures such as emergency stabilisation (straw mulching and seeding) and soil erosion barriers 588 are better at reducing soil erosion than natural regeneration or logging operations. However, post-fire measurements were not available and the model parameters were fixed based on literature. This is a 589 590 limitation of the present work. Further studies and field campaigns are needed to validate modelling results 591 and for adequately analysing the cost-effectiveness of these measures.

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#### 593 Credit Authors Statement

All co-authors conceptualized the study. AMDG designed the model simulations and wrote the initial draft. AMDG, RV, GFR, and TG collected input data. AMDG finalized the writing. All co-authors reviewed the paper. AMDG wrote the revised manuscript. RV, OC, AMDG and ALP secure funding and were involved in project administration.

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#### 599 Declaration of competing interest

600 The authors declare that they have no known competing financial interests or personal relationships that 601 could have appeared to influence the work reported in this paper.

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