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Effects of length and application rate of rice straw mulch on surface runoff and soil loss under laboratory simulated rainfall

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18 **Effects of length and application rate of rice straw mulch on surface runoff and**
19 **soil loss under laboratory simulated rainfall**

20

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42

43 **Abstract**

44

45 Forest land affected by deforestation yields high soil and water losses. Suitable
46 management practices need to be found that can reduce these losses and achieve
47 ecological and hydrological sustainability of the deforested areas. Mulch has been found
48 to be effective in reducing soil losses; straw mulch is easy to apply, contributes soil

49 organic matter, and is efficient since the day of application. However, the complex
50 effects of rice straw mulch with different application rates and lengths on surface runoff
51 and soil loss have not been clarified in depth. The current paper evaluates the efficiency
52 of rice straw mulch in reducing the hydrological response of a silty clay loam soil under
53 high intensity and low frequency rainfall events (tap water with total depth of 49 mm
54 and intensity of 98 mm/h) simulated in the laboratory. Surface runoff and soil loss at
55 three lengths of the straw (10, 30, and 200 mm) and three application rates (1, 2, and 3
56 Mg/ha) were measured in 50 (width) x 100 (length) x 10 (depth) cm plots with disturbed
57 soil samples (aggregate soil size < 4 mm) collected in a deforested area. Bare soil was
58 used as control experiment. Runoff volume and erosion were significantly (at $p < 0.05$)
59 lower in mulched soils compared to control plots. These reductions were ascribed to the
60 water absorption capacity of the rice straw and the protection cover of the mulch layer.
61 The minimum runoff was observed for a mulch layer of 3 Mg/ha of straw with a length
62 of 200 mm. The lowest soil losses were found with straw length of 10 mm. The models
63 developed predict runoff and erosion based on simple linear functions of mulch
64 application rate and length, and can be used for a suitable hydrological management of
65 soil. It is concluded that, thanks to rice straw mulch used as an organic soil conditioner,
66 soil erosion and surface runoff are significantly (at $p < 0.05$) reduced, and the mulch
67 protection contributes to reduce the risk of soil degradation. Further research is,
68 however, needed to analyze the upscaling of the hydrological effects of mulching from
69 the plot to the hillslope scale.

70

71 **Keywords:** Straw mulch; Soil erosion; Mulch application rate; Mulch length; Rainfall
72 simulator.

73

74 **1. Introduction**

75

76 Soil erosion is one of the most serious problems impacting the global environment
77 (Zhao et al., 2019). The impacts of soil erosion include land degradation, sedimentation,
78 and nutrient transport, resulting in reduced crop production, decay of soil properties,
79 and poor water quality (Pimentel et al., 1995). Inappropriate soil management practices
80 and land use generally cause these impacts on soils (Lucas-Borja et al., 2019;

81 Shabanpour et al., 2020), such as the increase of the erosion rates (Cherubin et al., 2017;
82 FAO, 2000).

83 Sustainable practices to control and mitigate soil erosion are essential worldwide and in
84 particular in the environments that are more prone to erosion risks. For instance,
85 deforestation removes the vegetal cover of woodlands, which usually protect the soil
86 surface from sealing and soil detachment. In the deforested environments, soil is left
87 bare and the lack of vegetation increases runoff and erosion rates.

88 A possible solution is the use of various types of inorganic mulch (e.g., gravel and other
89 soil particles) and organic mulch (e.g., crop residues) (Patil Shirish et al., 2013; Prats et
90 al., 2017). The term “mulch” refers to those materials - other than soil or living
91 vegetation - that function as a permanent or semi-permanent protective cover over the
92 soil surface (Jordán et al., 2011). Mulch protects the soil against raindrop impact,
93 reduces both the overland flow generation rates and velocity, allows improved
94 infiltration capacity and increases water intake and storage. These beneficial effects of
95 mulch noticeably reduce water and soil loss rates (Prosdocimi et al., 2016b).

96 The mulch types have variable levels of efficacy in controlling and mitigating soil
97 erosion and even in improving soil properties (de Lima et al., 2019). The increase in the
98 soil organic matter content can be particularly significant when vegetative residues are
99 used as mulch, as shown by García-Orenes et al. (2009) and Jordán et al.
100 (2010). Vegetal mulch types, such as leaf litter, cut-shrub barriers, wood-chips, crop
101 residues, and straw mulch (for instance, with rice or wheat) play, in general, an effective
102 influence on soil erosion rates (de Lima et al., 2019; Fernández et al., 2011; Jordan et
103 al., 2010). For example, in southern Spain Jordán et al. (2010) showed that a wheat
104 straw layer increases rain infiltration and delays runoff generation. In central China Liu
105 et al. (2012) showed that rice straw mulch significantly decreases the sediment yield.
106 Cerdà et al. (2016) showed the positive role of barley straw mulch to reduce the soil
107 erosion in persimmon plantations of eastern Spain. Prosdocimi et al. (2016a) found an
108 immediate reduction in soil losses in vineyards, when straw mulch was applied to soil.
109 However, some negative impacts of vegetal mulch on soil protection capacity have been
110 found in literature. For instance, compared to non-mulched soils, soil mulching with

111 straw or needle casts can increase erosion under heavy rainfall (Rahma et al., 2017;
112 Robichaud et al., 2013a, 2013b).

113 Rice, along with corn and wheat, is a common staple crop. The total harvested area of
114 rice is 160×10^6 ha globally, with most of the 700×10^6 t world production grown in
115 Asia (640×10^6 t) (IRRI, Africa Rice and CIAT, 2010; Hegde & Hegde, 2013). This
116 makes rice an important source of nutrition for Asia and, in general, worldwide. The
117 vegetal residues of rice cultivation (such as straw) are, therefore, abundant in several
118 countries and are becoming cheaper due to the decreasing demand for it as animal
119 fodder (Omidi-Mirzaee et al., 2017). Therefore, rice straw is a low-cost mulch substrate
120 to protect the soil and improve its fertility (Yadav et al., 2019). Rice straw can improve
121 the hydrological and physico-chemical properties of soil (Obour et al., 2019), thanks to
122 the incorporation into the soil of the ligno-cellulosic substances and the subsequent
123 degradation. Therefore, a practical use of rice straw mulch is beneficial for soil
124 conservation in deforested lands, which, as previously mentioned, are very susceptible
125 to land degradation of ecosystems once they lose the plant cover (Parhizkar et al.,
126 2020). Deforestation due to clear-cutting for timber production induces unsustainable
127 runoff generation and soil erosion rates. Therefore, it is important to evaluate whether
128 soil protection with rice straw mulch can be effective in controlling forest hydrology,
129 and the deforested lands of this country may represent a suitable case study.

130 In general, the influence of straw mulch on soil hydrology and biochemistry is well
131 documented in many studies worldwide, also for rice straw (Abrantes et al., 2018;
132 Fakhari et al., 2018; Gholami et al., 2013; Prats et al., 2017). However, it is believed
133 there are several factors influencing the effectiveness of straw mulch, including rice
134 variety, straw age and length, as well as application methods, rates, and seasons
135 (Mannering & Meyer, 1963; Pearson et al., 2015). The large number of these
136 influencing factors requires a better comprehension of the effects of rice straw mulch on
137 soil erosion, considering different rice straw characteristics as well as rainfall and soil
138 conditions.

139 At present, few studies have been done considering the effects of rice straw mulch
140 characteristics on runoff and soil loss (de Lima et al., 2019), particularly for a
141 deforested region. Recently, the latter authors found in a laboratory study that mulch
142 length affected soil loss more than runoff and that erosion decreased with the length of

143 rice straw applied to soil. Despite this isolated study, the need remains for a better
144 comprehension of the effects of rice straw mulch lengths and application rates on
145 erosion of deforested soils at high rainfall intensity. Laboratory studies using rainfall
146 simulators and soil plots under specific rain, soil, and vegetation factors are suggested in
147 order to control the effects of each factor influencing the erosion process (Bombino et
148 al., 2019).

149 To achieve these goals, the current study evaluates the hydrological effects (surface
150 runoff and soil loss) of three lengths (10, 30, and 200 mm) and three application rates
151 (1, 2, and 3 Mg/ha) of rice straw mulch on deforested soils using a rainfall simulator on
152 soil plots. The soil was sampled in a deforested hillslope of the Saravan Forest Park
153 (Northern Iran). It is hypothesized that the surface runoff and soil loss decrease with
154 higher length and application rate of rice straw. Finally, regression models are proposed
155 to predict runoff volume and soil loss from rice straw lengths and application rates.

156 The current research should give land managers insight about the most suitable soil
157 application method of rice straw in deforested areas, where the soil erosion rates are
158 high and the need for their reduction is compulsory, to avoid land degradation and other
159 negative environmental impacts.

160

161 **2. Materials and methods**

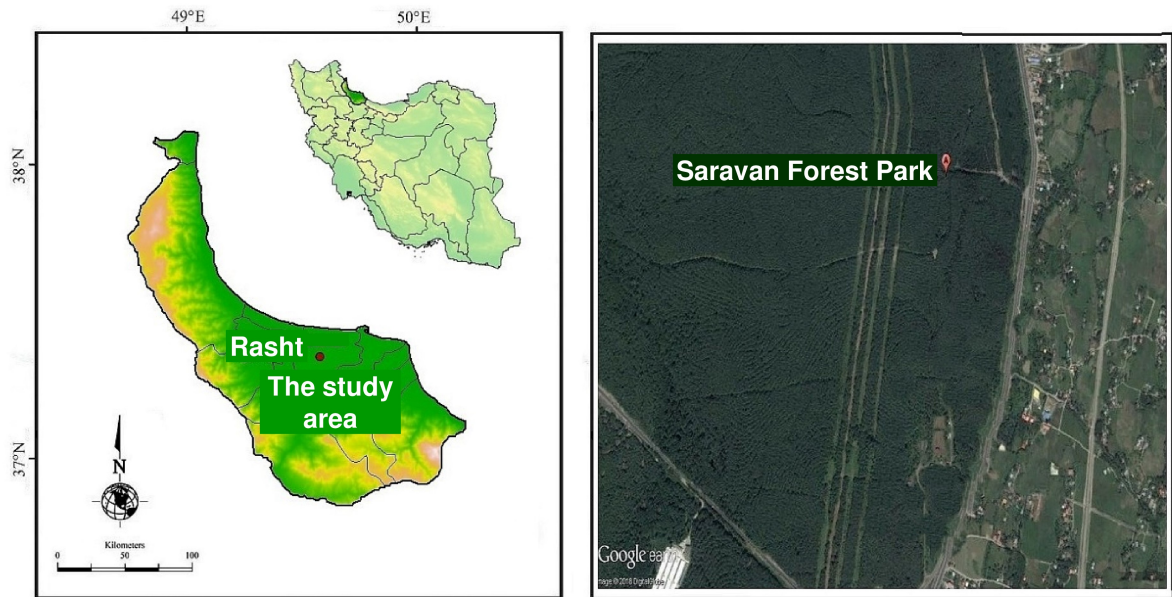
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163 *2.1. Soil sampling, analysis, and characterisation*

164

165 In Iran, deforestation is one of the most important anthropogenic factors of soil
166 degradation and erosion, especially in the northern part of the country, where
167 deforestation due to illegal logging is one of the major factors causing severe soil
168 erosion (Bahrami et al., 2010; Emadodin, 2008). The soil for the laboratory experiment
169 was selected from a deforested hillslope of the Saravan Forest Park, which is one of the
170 oldest forestlands in Guilan province. The park is located in the south of Rasht city and
171 the outlet coordinates are 37°08'04" N and 49°39'44" E (Fig. 1).

172



173
174

175 **Fig. 1.** Geographical location and aerial map (source: Google[®] Maps[®]) (Saravan Forest
176 Park, Guilan province, northern Iran).

177
178

179 The Saravan Forest Park is located at a mean altitude 93 m a.s.l. with the slope gradient
180 varying from 12 to 25%. Some hillslopes inside the park, which were deforested to
181 install high-voltage towers one to three years before the investigation, were previously
182 covered by different plant species (trees, shrubs, and herbs) with the highest density
183 among all hillslopes in the park (Parhizkar et al., 2020). The plant biodiversity of the
184 Saravan Forest Park is ample. Some dominant species include *Carpinus betulus*,
185 *Quercus castaneifolia*, *Pinus taeda* and *Parrotia persica*.

186 According to the Köppen-Geiger classification, the area is characterized by a typical
187 Mediterranean climate, *Csa* type (Kottek et al., 2006). The mean annual temperature
188 and precipitation are 16.3°C and 1360 mm, respectively (IRIMO, 2016).

189 Soil samples were randomly collected from the top layer (0 to 50 cm) of the deforested
190 hillslope (Kukul & Sarkar, 2010), using the procedure suggested by Singh Sidhu (2015).
191 Before sampling, weeds, rocks, and litter were removed from the soil surface. Then, the
192 soil was transported to the Soil Testing Laboratory of the College of Agriculture, Guilan
193 University. The soil samples were sieved through a 4-mm mesh, to remove the residual

194 gravel and vegetation, and then well mixed. Here, the soil was maintained under a
195 tarpaulin cover until the experiment date, when it was placed in the experimental plots
196 (see section 2.2).

197 The soil texture was silty clay loam (SDSD, 2017) and the aggregate stability in water,
198 bulk density, and organic matter content of the soil were measured on representative
199 sub-samples of the collected soil samples. Sand, silt, and clay contents of the soils were
200 measured by sieving and hydrometers. Bulk density and aggregate stability were
201 determined using the oven-drying and the wet-sieving methods, respectively. Soil
202 organic matter was estimated using the potassium dichromate colorimetric method.

203

204 *2.2. Soil characteristics*

205

206 The mean clay, silt, and sand contents of the studied soil were $37.5 \pm 0.02\%$ (where the
207 \pm is the standard deviation), $49.9 \pm 0.01\%$, and $12.6 \pm 0.01\%$, respectively. The bulk
208 density was $1487 \pm 38 \text{ kg/m}^3$, while the soil aggregate stability, a main indicator of the
209 ability of soil aggregates to resist degradation, was 0.21 ± 0.03 . The soil aggregate
210 stability is lower compared to the reference values (0.70-0.75, Soil Quality Institute,
211 1998) and those measured by Parhizkar et al. (2020) in the same area (Guilan province,
212 0.25-0.66), who always reported a large variability of this parameter.

213 The sampled soil had a mean organic matter content of $1.22 \pm 0.05\%$, which is lower
214 compared the contents (from 2.8 to 3.4%) measured in croplands and gardens in the
215 same area (Guilan province) by Shabanpour et al. (2020), but similar to the values (from
216 1.28 to 1.87%) reported by Parhizkar et al. (2020) in woodland and forestland of the
217 same park.

218

219 *2.3. Plot description*

220

221 The experimental plots consisted of timber planks (0.5-m wide, and 1-m long with 0.1-
222 m high sides) (Fig. 2a), placed on concrete blocks at a slope of 12% (Shoemaker, 2009;
223 Singh Sidhu, 2015). The base of each plot was made of wood, which was not
224 impervious to water. Small holes were drilled in the base, in order to facilitate water
225 drainage and avoid unrealistic saturation of the soil.

226 Before the experiments, the soil was air-dried until optimal water content, in order to
227 maintain the stability of soil aggregates (Kukul & Sarkar, 2010). Then, the soil was
228 placed in the plots and the surface was gently leveled by hand. A tarpaulin cover was
229 put on the top, in order to avoid water evaporation from the plot. The plot was equipped
230 with a horizontal collector placed at the downstream side, which conveyed the flows of
231 water and sediment into a plastic tank through a PVC pipe.

232

233 *2.4. Rainfall simulator*

234

235 Runoff volume and soil loss were measured between June and July 2019, when rain was
236 simulated on the plot using a hand-crafted simulator (Fig. 2b). The rainfall simulator
237 consisted of two open rectangular boxes, whose bottom was made of a squared grid.
238 The grid was equipped with 70 syringe needles with a diameter of 2.5 mm. The syringe
239 needles, with an outer diameter of 0.7 mm and a length of 40 mm, were uniformly
240 installed 3.1 m above the ground, to provide a rectangular 0.5 m x 1 m spray area. Drop-
241 former rainfall simulators are widely used in the laboratory due to their accuracy.

242 The rainfall intensity was controlled by feeding the boxes with a flow of tap water
243 (drawn from the municipal aqueduct). This flow was kept constant throughout the
244 experiment via a pipe. Before starting the experiment, the rainfall simulator was
245 calibrated at a rainfall intensity of 98 ± 1.1 mm/h. The experiment was set to this very
246 high value, since extreme weather conditions result in the highest erosion rates in this
247 area. In more detail, the Rasht area has an annual mean rainfall depth of 1353 ± 279 mm
248 with historical (years 1951-2003) extremes of more than 2000 mm (Modarres, 2006;
249 Rahimzadeh et al., 2009). Considering that the climate is typically Mediterranean,
250 where few rainfall events (often two to five) lasting one to two hours account for half of
251 the total precipitation (Modarres, 2006), an intensity of 90-100 mm/h¹ is realistic, and
252 this may result in very erosive precipitations.

253 The walls of the laboratory prevented wind from disturbing the simulated rain.
254 However, the plots were exposed to a moderate air stream that slightly varied the impact
255 positions of the falling drops. The distribution uniformity of the rainfall intensity (Duke
256 & Perry, 2006) was 83%, a value that can be considered as good in the classification of
257 The Irrigation Association (2002).

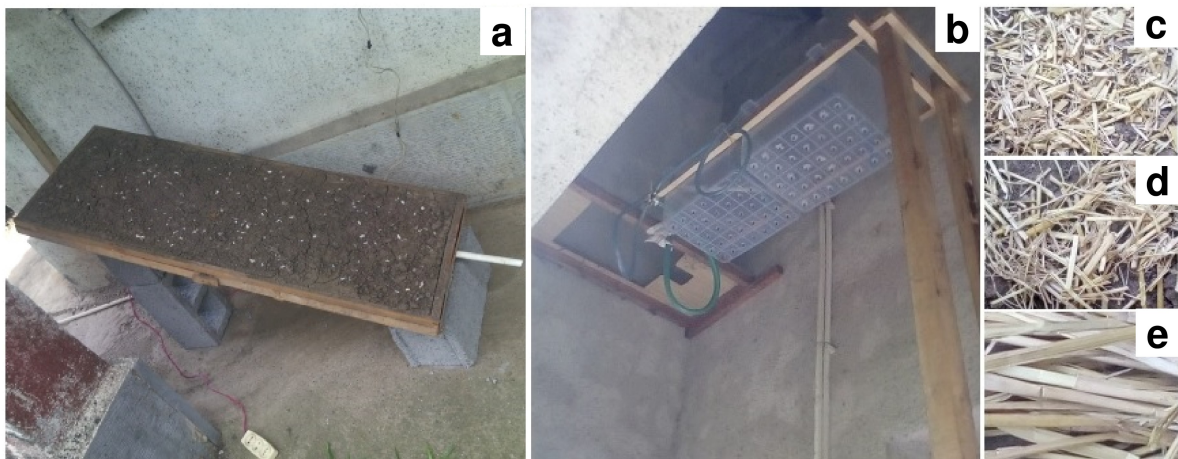
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259 2.5. Straw mulch characteristics

260

261 The rice variety *Oryza sativa* L. was used for the experimental straw mulch. This
262 variety is considered as one of the most important cultivated rice species in the
263 agricultural fields of northern Iran (Yousefian et al., 2019). Three lengths (10, 30, and
264 200 mm) of the rice straw mulch were used, as suggested by de Lima et al. (2019). The
265 200-mm straw length was obtained by breaking the straw particles by hand, whereas the
266 other lengths were produced by shredding (30 mm) and rice grain husking machines (10
267 mm). A uniform cover of straw mulch was applied over the entire soil surface of the
268 plot for each length (Fig. 2c-e).

269



270

271 **Fig. 2.** The experimental plot (a); rainfall simulator (b); 10-mm length rice straw mulch
272 (c); 30-mm length rice straw mulch (d); 200 mm-length rice straw mulch (e), used for
273 the experiment.

274

275

276 A “water absorption capacity” of mulch was estimated on a sample of 100 g of each
277 length (dry weight). This sample was placed on the soil of the plot and a rainfall
278 intensity of 95 mm/h¹ was simulated for 30 min. The water absorption capacity (WAC,
279 %) was:

$$WAC = \frac{w_w - w_d}{w_d} \times 100 \quad (1)$$

280

281 where w_w and w_d (g) are the sample weights after and before rainfall, respectively. The
282 wet straw was immediately weighed, in order to limit the water and soil losses.

283 The estimated values of WAC for 10, 30, and 200-mm lengths of rice straw mulch were
284 30, 52, and 82%, respectively. Finally, surface cover of soil due to straw mulch
285 application was measured by photographic method followed by image processing using
286 common software.

287

288 *2.6. The experimental design*

289

290 Before the tests, the soil was saturated with tap water until ponding. Water was gently
291 and slowly poured on the plot surface to avoid runoff, splashing, and slaking. Then, the
292 soil was left to dry in the open air for 24 hours, to have a water content equal to the field
293 capacity. For each experiment, a 5-10 mm layer of was removed from the plot surface
294 and replaced by a new layer of dry soil, in order to ensure the same content of soil
295 particles. To avoid discontinuities between the old lower and fresh upper layers, the
296 upper surface of the lower layer was roughened using a manual ripper. After preparing
297 the soil with the desired straw mulch application rate and length and filling the rainfall
298 simulator with water, the experiment started, and the runoff volume and soil loss were
299 collected and measured.

300 An experiment with bare soil in the plot was considered as the “control”. For the other
301 experimental runs, three application rates (1, 2, and 3 Mg/ha) and three lengths (10, 30,
302 and 200 mm) of rice straw mulch were tested (after de Lima et al., 2019). The weight of
303 rice straw mulch for 1, 2, and 3 Mg/ha application rates was 71, 142, and 213 g,
304 respectively. Each test was done in triplicate. Therefore, 30 experiments were done (3
305 application rates x 3 lengths x 3 replicates + 1 control x 3 replicates). Each experiment
306 was done for 30 min as the runoff discharge was stable in all the experiments by that
307 time (Zhao et al., 2019). After measuring the runoff volume, the collected water was
308 oven-dried at 80°C for 24 h, to measure the sediment weight. Moreover, the runoff
309 outlet time (the time when runoff water starts to drop in the collecting tanks) was
310 measured. This time gives information about the connectivity within the plot.

311 Hereinafter, each experiment will be indicated as “ARXX-LXXX”, where “ARXX” is
312 related to the mulch application rate and “LXXX” to the mulch length. For instance,
313 AR1-L30 indicated the plots covered by 1 Mg/ha of straw with a length of 30 mm.

314

315 *2.7. Statistical analysis*

316

317 Using QQ-normal plots, the normal distribution hypothesis of the samples was checked.
318 An ANalysis Of VAriance (ANOVA) was used to assess the statistical significance of
319 the differences in the runoff volume and soil loss (considered as the dependent
320 variables) among the different straw mulch application rates and lengths (independent
321 variables). Then, a Principal Component Analysis (PCA) was applied, in order to find
322 correlations (using Pearson’s method) among runoff, soil loss, and mulch application
323 rate, length, and cover, as well as to identify the existence of meaningful derivative
324 variables (Principal Components, PCs) (Rodgers & Nicewander, 1988). The
325 correlations between runoff volume and soil loss (dependent variables), and mulch rate
326 and application rate (independent variables) were analyzed by linear multi-regression
327 equations. The simulations were evaluated for “goodness-of-fit” with the corresponding
328 observations. First, observed and simulated values of the water flow were visually
329 compared in scatterplots. Then, the following indicators, commonly used in the
330 hydrological literature (e.g., Legates & McCabe, 1999; Loague & Green, 1991;
331 Willmott, 1982), were applied for a quantitative evaluation: (i) the main statistics (i.e.,
332 the maximum, minimum, mean, and standard deviation of both the observed and
333 simulated values); (ii) a set of summary and difference measures, such as the coefficient
334 of determination (R^2), coefficient of efficiency (E), and its modified form (E*, Willmott,
335 1982), and Root Mean Square Error (RMSE). In particular, E is more sensitive to
336 extreme values, while E* is better suited to significant over- or underprediction by
337 reducing the effect of squared terms. The related equations are reported in Zema et al.
338 (2012), Krause et al. (2005), Moriasi et al. (2007), and Van Liew & Garbrecht (2003).

339 To summarize:

340 - R^2 ranges from 0 (no agreement between model and data variance) to 1 (perfect
341 agreement); values over 0.5 are acceptable (Santhi et al., 2001; Van Liew et al.,
342 2003; Vieira et al., 2018);

343 - E (Nash & Sutcliffe, 1970) and E^* are the most common measure of model
344 accuracy and range from $-\infty$ to 1; the model accuracy is “good” if E and $E^* \geq 0.75$,
345 “satisfactory” if $0.36 \leq E$ and $E^* \leq 0.75$, and “unsatisfactory” if E and $E^* \leq 0.36$
346 (Van Liew & Garbrecht, 2003);

347 - RMSE, which measures the standard deviation between observations and
348 predictions, should be as close as possible to zero (Fernandez et al., 2010); RMSE is
349 considered good if its predicted value is lower than 0.5 of the observed standard
350 deviation (Singh et al., 2004).

351 All statistical analyses were done with the SPSS 17.0 and XLSTAT 9.0 software.

352

353 **3. Results**

354

355 *3.1. Analysis of the hydrological variables*

356

357 Table 1 lists the volumes and outlet times of runoff as well as the soil losses measured
358 in the experimental plots under the various rice straw mulch lengths and application
359 rates. The control plot produced the highest runoff volume (13.2 ± 0.23 mm), while the
360 lowest value was observed in the AR3-L200 plots (7.62 ± 0.12 mm) (Table 1).

361 In general, for a given application rate of mulch, the runoff volume decreased and the
362 outlet time increased when the straw length increased. The same trend (decreasing
363 volume and increasing time) can be noticed, if the application rate of mulch increases at
364 the same straw length (Table 1).

365 This is better explained in Fig. 3a, where it can be noticed that, if the runoff volumes are
366 averaged among the plots with the same mulch length, but different application rates, a
367 significant ($p < 0.05$) decreasing trend for runoff with increasing application rate is
368 evident (11.31 ± 0.10 mm in AR1, to 8.49 ± 0.05 mm in AR3). Conversely, comparing
369 plots with the same mulch application rate, but different lengths, runoff decreased
370 (significantly for the finer straw lengths, $p < 0.05$) when the length increased (from
371 10.67 ± 0.12 mm in L10 to 9.04 ± 0.06 mm in L200) (Fig. 3a). The lowest runoff outlet time
372 was found in the control plot (49 s) and the highest in AR3-L200 plots (122 s) (Table
373 1).

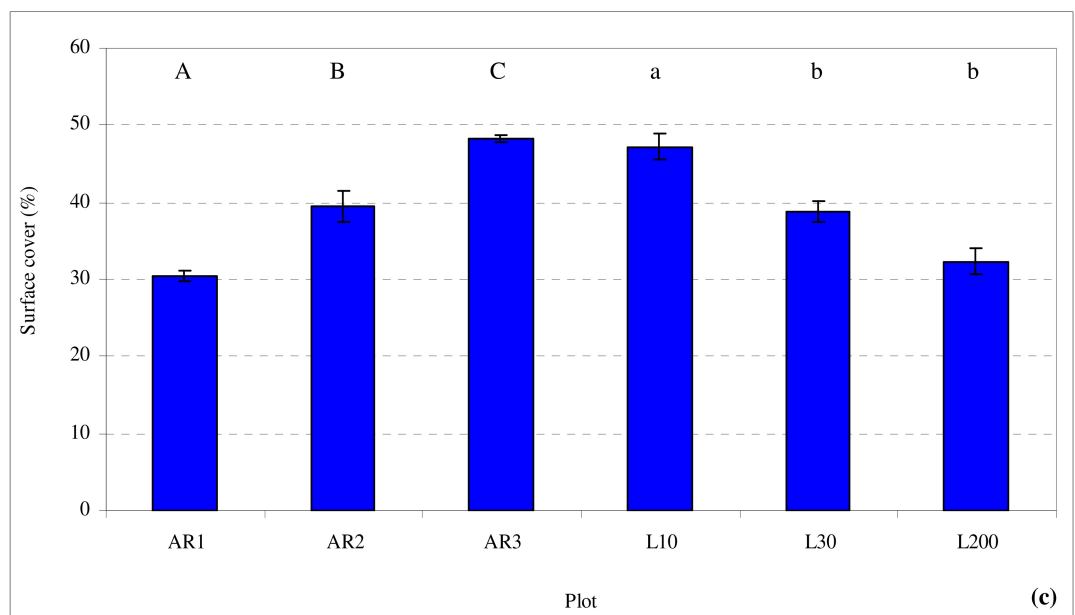
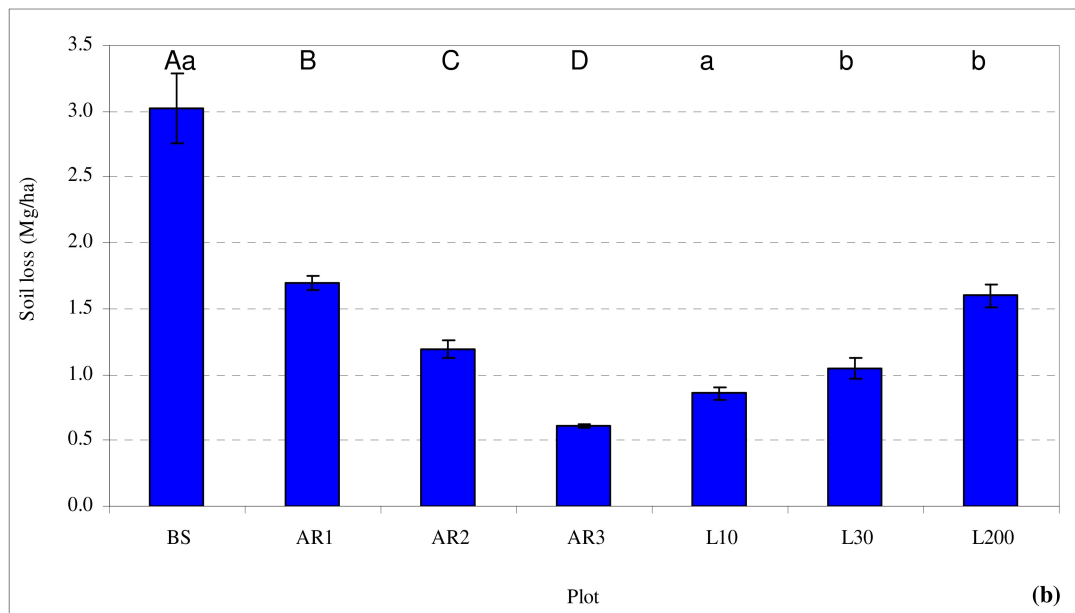
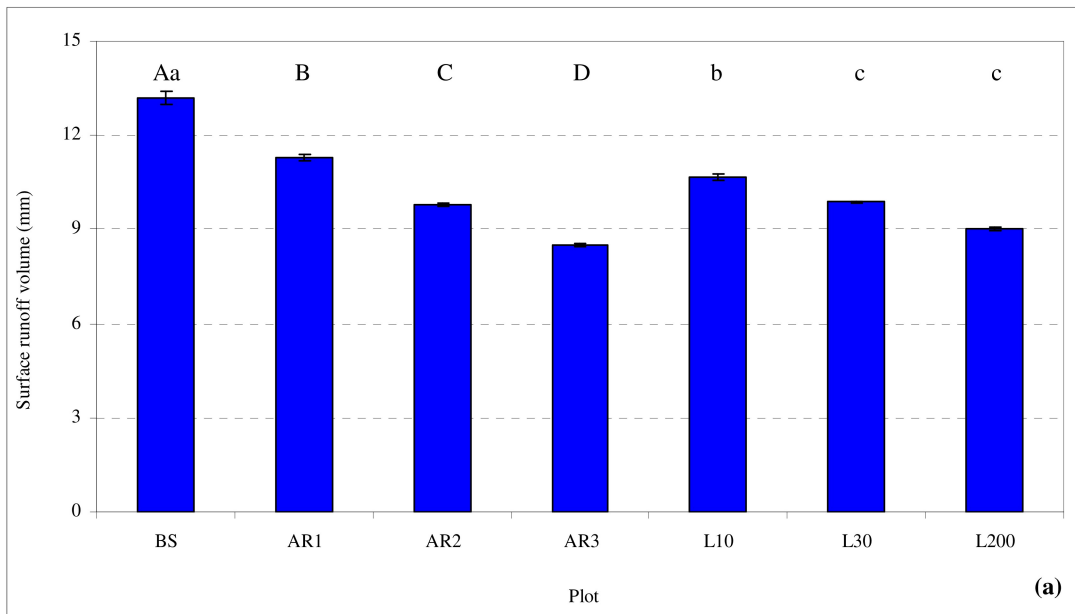
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375 **Table 1.** Experimental conditions (mulch application rate and length), and surface
 376 cover, runoff volume, runoff outlet time, and soil loss evaluated at the plot scale on a
 377 deforested soil sampled from the Saravan Forest Park (northern Iran).
 378

Plot	Mulch characteristics		Surface cover (%)	Runoff volume (mm)	Runoff outlet time (s)	Soil loss (Mg/ha)
	Application rate (Mg/ha)	Length (mm)				
BS	0 (bare soil)	-	-	13.20 ± 0.23	49	3.02 ± 0.26
AR1-L10	1	10	38.9 ± 2.1	12.20 ± 0.28	60	1.33 ± 0.10
AR1-L30		30	27.7 ± 1.1	11.52 ± 0.12	68	1.48 ± 0.16
AR1-L200		200	24.6 ± 1.2	10.21 ± 0.10	79	2.28 ± 0.20
AR2-L10	2	10	47.8 ± 5.0	10.39 ± 0.10	84	0.82 ± 0.01
AR2-L30		30	39.9 ± 2.6	9.66 ± 0.08	89	1.11 ± 0.02
AR2-L200		200	30.8 ± 1.3	9.29 ± 0.02	100	1.65 ± 0.12
AR3-L10	3	10	54.8 ± 3.2	9.42 ± 0.05	107	0.42 ± 0.04
AR3-L30		30	48.7 ± 3.8	8.44 ± 0.14	113	0.55 ± 0.02
AR3-L200		200	41.4 ± 4.1	7.62 ± 0.12	122	0.87 ± 0.02

379 Note: BS = bare soil; in the plot indications (“ARXX-LXXX”), “ARXX” is related to the mulch application rate, and
 380 “LXXX” to the mulch length.

381



383

384 **Fig. 3.** Total runoff volume (a), soil loss (b), and surface cover (c) averaged among
385 application rates and lengths of straw mulch applied to a deforested soil and evaluated at
386 the plot scale on a deforested soil sampled from the Saravan Forest Park (northern Iran).

387

388 Note: Different lowercase and capital letters indicate significant differences among mulch sizes and doses
389 at p-level < 0.05; BS = bare soil; in the plot indications (“ARXX-LXXX”), “ARXX” is related to the mulch
390 application rate and “LXXX” to the mulch length. The vertical lines on the bars indicate the standard
391 deviations.

392

393 Soil erosion was maximum for the bare plot (3.02 ± 0.2 Mg/ha). The lowest erosion was
394 measured in AR3-L10 plots (0.42 ± 0.04 Mg/ha) (Table 2). It is also interesting to note
395 that a high soil loss (2.28 ± 0.20 Mg/ha) was detected in the deforested soil (plots AR1-
396 L200) treated with 1 Mg/ha of 200-mm rice straw, but this value is lower by about 25%
397 compared to the bare soil, showing how mulching with an unsuitable dose and length is
398 still able to significantly reduce soil erosion.

399 As noticed for the runoff, for a given straw length, the soil loss decreased when the
400 mulch dose increased. Instead, and differently from what observed for runoff, erosion
401 increased if the application rate was kept constant, but the straw length was increased
402 (Table 1). These trends are evident observing Fig. 3b, which shows that, under the same
403 mulch length, soil loss significantly ($p < 0.05$) decreased with increasing mulch rate
404 (from 1.70 ± 0.05 Mg/ha in AR1 to 0.61 ± 0.01 Mg/ha in AR3). Conversely, as the
405 mulch length decreased under a constant application rate, soil loss increased ($0.86 \pm$
406 0.04 Mg/ha in L10 to 1.60 ± 0.09 Mg/ha in L300), but the differences were significant
407 ($p < 0.05$) only between BS and L10 on one side and L30 and L200 on the other side
408 (Fig. 3b).

409 Comparing the plots with straw mulch application, the lowest and the highest surface
410 cover were measured in AR1-L200 plots ($24.6 \pm 1.06\%$) and AR3-L10 ($54.8 \pm 3.2\%$),
411 respectively (Table 1). The variability of surface cover was the opposite of the soil
412 erosion trend among mulch length and application rate, as shown by Fig. 3c, in which
413 the values of surface cover are averaged among the different mulch application rates
414 and lengths. In other words, surface cover increased with the mulch application rate

415 (from $30.4 \pm 0.62\%$ in AR1 to $48.3 \pm 0.46\%$ in AR3) and decreased with its length (44.5
 416 $\pm 0.53\%$ in L10 to $32.3 \pm 1.69\%$ in L200) under the same length or application rate,
 417 respectively. The differences in surface cover were always significant ($p < 0.05$) at
 418 different mulch application rates; instead, the length L30 was significantly ($p < 0.05$)
 419 different from L10, but not from L200 (Table 1).

420

421 3.2. Analysis of relations between the hydrological variables and the mulch parameters

422

423 The analysis of Pearson's matrix shows a positive correlation between total runoff on
 424 one side, and soil loss ($r = 0.66$) and straw length ($r = 0.91$). Moreover, runoff was
 425 negatively correlated with surface cover ($r = -0.65$) as well as mulch application rate (r
 426 $= -0.51$). Soil loss also was negatively correlated surface cover ($r = -0.95$) and mulch
 427 application rate ($r = -0.87$), but not with mulch length ($r = 0.16$) (Table 2).

428

429 **Table 2.** Pearson's correlation matrix among the hydrological variables and mulch
 430 characteristics in plots treated with three lengths and three application rates of rice straw
 431 mulch applied to a deforested soil sampled from the Saravan Forest Park (northern
 432 Iran).

433

Variables	Mulch application rate	Mulch length	Surface cover	Runoff volume	Soil loss
Mulch application rate	1	0.174	0.842	-0.909	-0.872
Mulch length		1	-0.099	-0.516	0.162
Surface cover			1	-0.649	-0.948
Runoff volume				1	0.663
Soil loss					1

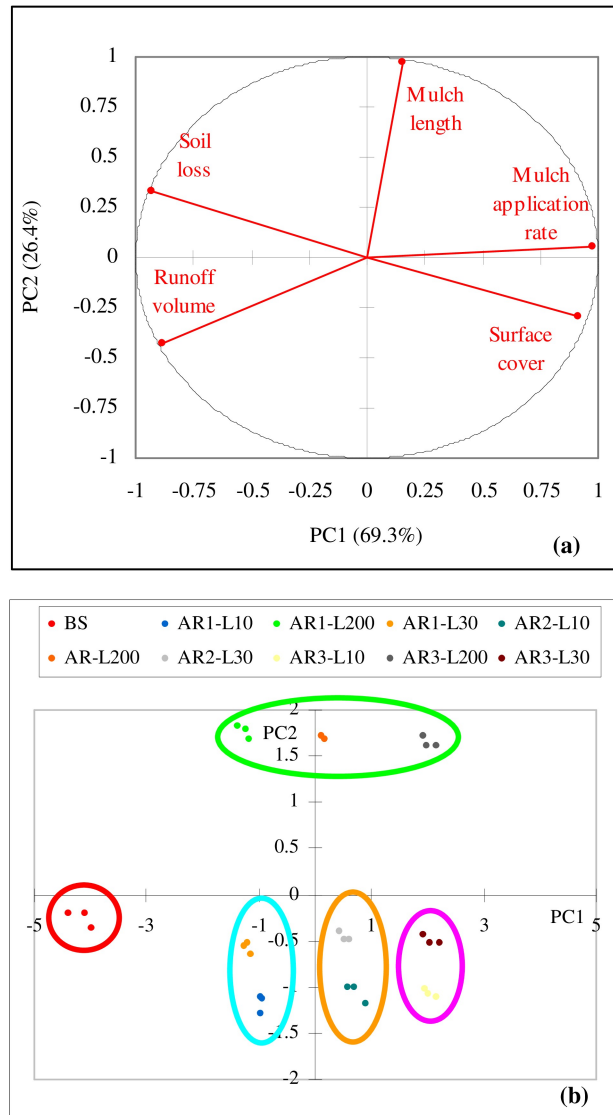
434

Note: Values in bold are significant at p level < 0.05 .

435

436 Two principal components (PCs) were identified using PCA, and explained together
 437 97% of the total variance of the hydrological variables and straw mulch parameters
 438 (69% for PC1 and 26% for PC2).

439 The mulch application rate and surface cover as well as runoff and soil loss had high
 440 (absolute value > 0.88) positive and negative loadings, respectively, on PC1, while only
 441 mulch length significantly ($p < 0.05$) influenced PC2 (loading over 0.97) (Fig. 4a). In
 442 other words, runoff and soil loss were associated with low values of the mulch
 443 application rate and surface cover (Fig. 4b).



444

445

446 **Fig. 4.** Loadings of the original hydrological variables and straw mulch characteristics
 447 (length, application rate, and surface cover) (PC₁ and PC₂) (a) and scores on the first
 448 two Principal Components provided by PCA applied to plots (b) with deforested soils
 449 sampled from the Saravan Forest Park (northern Iran).

450

451 Note: BS = bare soil; in the plot indications (“ARXX-LXXX”), “ARXX” is related to the mulch
452 application rate and “LXXX” to the mulch length.

453

454 Plotting the hydrological variables and the associated mulch parameters on the two PCs,
455 five well differentiated clusters were evident: a first cluster grouping the control plots
456 (associated with low values of PC1), a second group with AR1-L200, AR2-L200 and
457 AR3-L200 plots, associated with high values of PC2) and four other clusters with the
458 remaining plots, characterized by intermediate values of PC1 and low values of PC2
459 (Fig. 4b).

460

461 3.3. Modeling runoff volume and soil loss using mulch parameters

462

463 Table 3 lists the coefficients of the equations estimating runoff volume and soil loss
464 from mulch application rate and length.

465

466 **Table 3.** Coefficients of the multi-regression equations between runoff volume or soil
467 loss and straw mulch parameters (application rate, [Mg/ha]), and length, [mm]) in plots
468 treated with different lengths and application rates of straw mulch applied to a
469 deforested soil sampled from the Saravan Forest Park (northern Iran).

470

Model parameter	Runoff volume	Soil loss
Intercept	13.275	2.506
Mulch application rate	-1.429	-0.760
Mulch length	-0.008	0.002
Mulch application rate x length	0.001	0.001

471

472 The proposed equations are the following:

473

$$RV = -1.429 MAR - 0.008 ML + 0.001 ML \cdot MAR + 13.275 \quad (2)$$

$$SL = - 0.760 MAR + 0.002 ML + 0.001 ML \cdot MAR + 2.506 \quad (3)$$

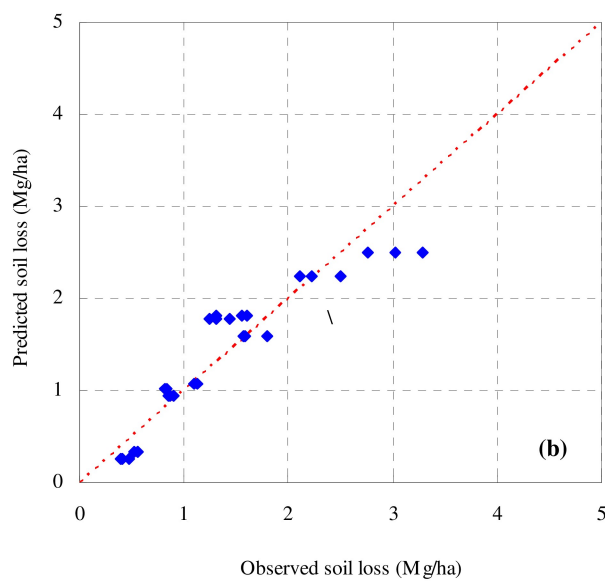
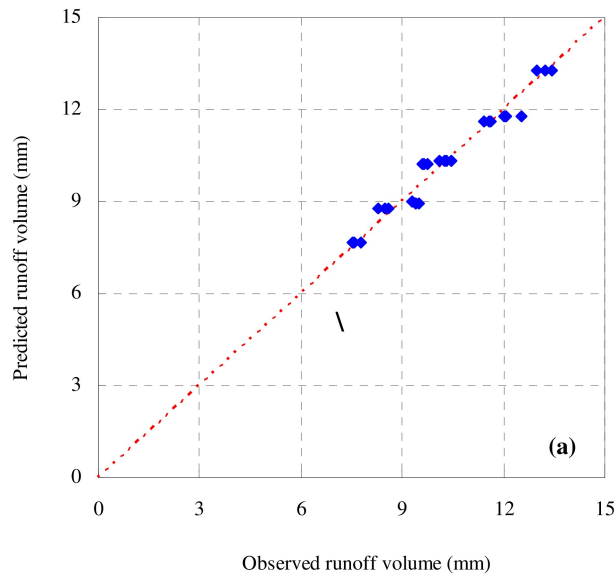
474

475 where RV = surface runoff volume (mm), SL = soil loss (Mg/ha), MAR = mulch
 476 application rate (Mg/ha), and ML = mulch length (mm).

477

478 The explanatory capacity of these equations was very high for both the modeled
 479 hydrological variables (R^2 equal to 0.96 for surface runoff and 0.87 for soil loss). The
 480 predictions of both surface runoff and soil loss were very close to the line of perfect
 481 agreement (Fig. 5).

482



483

484

485 **Fig. 5.** Scatterplots of runoff volume (a) or soil loss (b) observed and predicted using
486 the multiregression models based on rice straw mulch parameters (application rate and
487 length) in plots with a deforested soil sampled from the Saravan Forest Park (northern
488 Iran).

489

490

491 Not only are the statistics of the observed and predicted variables very close (maximum
492 difference of 33.5% for the maximum values of soil loss), but also the indexes gave
493 values exceeding the acceptance limits suggested by the literature (Santhi et al., 2001;
494 Singh et al., 2004; Van Liew et al., 2003; Vieira et al., 2018; Van Liew & Garbrecht,
495 2003). In more detail, E was good for runoff and soil loss (0.96 and 0.87, respectively),
496 while E* was good for runoff (0.80) and satisfactory (0.65) for soil loss. The values of
497 RMSE were always lower than 50% of the observed standard deviations (Table 4).

498

499 **Table 4.** Values of the criteria adopted for evaluating the accuracy of equations (2) and
500 (3) to predict the soil loss and runoff volume from mulch parameters in plots treated
501 with different lengths and application rates of straw mulch applied to deforested soils
502 sampled from the Saravan Forest Park (northern Iran).

503

Hydrological variable		Statistic				Index			
		Mean	Min	Max	Std. Dev.	R ²	E	E*	RMSE
Runoff volume	Observed	10.2	7.5	13.4	1.7	0.96	0.96	0.80	0.33
	Predicted	10.2	7.7	13.3	1.6				
Soil loss	Observed	1.35	0.39	3.29	0.78	0.87	0.87	0.65	0.28
	Predicted	1.35	0.26	2.51	0.73				

504 Note: Min = minimum; Max = maximum; Std. Dev. = Standard Deviation; R² = coefficient of determination; E and
505 E* = coefficients of efficiency of Nash and Sutcliffe (1970) in the original (E) and modified (E*) form; and RMSE =
506 root mean square error (expressed in mm for runoff volume and Mg/ha for soil loss).

507

508

509 4. Discussion

510

511 *4.1. The influence of mulching conditions on runoff volume and soil loss*

512

513 Previous studies have evaluated how much straw influences the hydrological response
514 of the soil under different experimental conditions (e.g., de Lima et al., 2019; Gholami
515 et al., 2013, 2014; Keesstra et al., 2019; Lucas-Borja et al., 2018; Sadeghi et al., 2015).
516 However, the research done in the field is highly affected by other factors, such as the
517 rainfall intensity, spatial variability of soil properties, plant cover, and soil moisture. In
518 the current study, the straw mulch cover has been isolated to assess its effect through
519 controlled experiments in the laboratory. Therefore, the effects of mulch application
520 rates and lengths on the variability of the soil loss and runoff volume can be directly
521 evaluated at the plot scale.

522 The presence of straw mulch reduced by 8% (plots AR1-L10) to 42% (plots AR3-L200)
523 the runoff volume and by 25% (plots AR1-L200) to 86% (plots AR3-L10) the soil
524 erosion rate. The lower runoff volumes in the straw-mulched experiments compared to
525 bare soil (control plots) are consistent with findings of several authors (e.g., Adams,
526 1966; ; Gholami et al., 2013; Liu et al., 2012). In every case, mulching soil with straw,
527 also with low application rates and coarse sizes, is beneficial for improving the
528 hydrological response of deforested soils, since the current study has demonstrated that
529 runoff decreases at least by 7-10% and soil erosion by 25% or much more. These
530 positive effects on soil hydrology support other hydrological and ecological advantages,
531 such as the increase in water capacity retention and infiltrability as well as the
532 improvement of some important physico-chemical properties (Prosdocimi et al., 2016b),
533 which, however, go beyond the specific aims of the current paper. The current study
534 confirms the immediate impact of straw mulch to reduce the runoff generation capacity
535 and erosion of soils, such as Prosdocimi et al. (2016a) found in field experiments in
536 eastern Spain under vineyard cultivation. Surface runoff and soil loss decrease in
537 mulched soils due to three main factors. First, straw mulch has a capacity to absorb
538 water (from 30 to 82% of the precipitation, depending on the mulch length). This water
539 volume is retained by the straw, reducing the runoff volume. Second, the presence of
540 straw over the soil represents an obstacle against the overland flow, which decreases the

541 flow velocity. Third, the mulch layer protects the soil surface against raindrop impact,
542 acting as a protection against the precipitation erosivity.

543 The significant capacity of straw to absorb water is beneficial, since the mulching layer
544 decreases the share of precipitation that turns into runoff, and, therefore, the detachment
545 capacity of the overland flow.

546 The decrease in the flow velocity due to the presence of straw over the soil is
547 demonstrated by the reduction of the runoff outlet time (the lowest in the control plot
548 and the highest in the AR3-L200 plots), which increases upon mulch length and
549 application rate. This reduction is in accordance with findings of many authors (e.g., de
550 Lima et al., 2019; Keesstra et al., 2019; Yanosek et al., 2006), who concluded that straw
551 mulch is effective in delaying the runoff outlet time or runoff initiation. It is also
552 important to note that, when the mulch application rate and length increase, the runoff
553 generation capacity significantly ($p < 0.05$) decreases and then the runoff outlet time is
554 delayed. Therefore, an application rate of 3 Mg/ha with a length of 200 mm is suggested
555 for the highest runoff reduction. These results are consistent with those of de Lima et al.
556 (2019), who found that 10-mm mulch yielded the highest runoff.

557 The protection effect of straw against the precipitation erosivity helps to reduce the
558 hydrological response of mulched soil, reducing erosion. The mulch layer protects the
559 soil surface against raindrop impact, which is one of processes determining erosion, in
560 addition to the transport capacity of runoff. In the current study, the lowest erosion was
561 detected for the AR3-L10 plots, that is, in the plots with the highest mulch application
562 rate (as for surface runoff), but the lower length. This lowest soil loss may be due to the
563 fact that these mulch conditions lead to the highest surface cover, and, thus, the
564 maximum soil protection. The reduced erosion with the lower surface runoff and the
565 higher soil protection due to mulch characteristics are also confirmed by the positive
566 correlations between total runoff, soil loss, and mulch application rate and the negative
567 relations with surface cover as well as straw length. In other words, runoff and soil loss
568 are associated with low values of the mulch application rate and surface cover.

569 The two smaller lengths of rice straw mulch (10 and 30 mm) present much more
570 complex pathways for runoff. These pathways should enhance deposition of suspended
571 sediments to be deposited when the flow rates decrease, while the overland flow was

572 not influenced. In the case of the 200-mm straw, the mulch seems to increase soil
573 erosion due to the straighter pathways. This is consistent with Rahma et al. (2017), who
574 reported that straw mulch can induce greater soil losses compared to non-mulched soils
575 under extreme rainfall conditions, such as those of the current study. As a matter of fact,
576 the longer straw length resulted in greater soil losses, because the straw layer provides
577 straighter pathways that can accelerate flow velocity and concentrate surface flow. This
578 effect should be considered with caution when the straw length must be identified for
579 mulching, and crushing the straw as fine as possible before land spreading for soil
580 protection should be done.

581 It is interesting to note that soil erosion is not directly dependent on mulch length (that
582 is, there is not a clear trend in soil loss reduction with straw size), but only to mulch
583 application rate, which influences surface cover. This is confirmed by PCA, which
584 shows direct associations among four of the five variables analyzed (runoff, soil loss,
585 surface cover, and mulch application rate) and the first PC (which can be considered a
586 synthetic measure of the soil hydrological response). The latter, in turn, is weakly
587 associated with straw length. Moreover, the evident clustering of experiments provided
588 by PCA clearly associate causes (length and application rate of straw mulch, and surface
589 cover) and effects (runoff and soil loss). The very high correlations between the
590 hydrological variables measured in the current study and the mulch application rate
591 indicate that the latter is the factor with the greatest influence on the hydrological
592 response of a deforested soil, while mulch length is more important for runoff reduction
593 than for erosion control. For this purpose, rice straw application is beneficial to increase
594 the surface cover, which is very effective to reduce soil loss, as shown by the high
595 correlation between these two variables. As regards in particular the experiments done
596 using rice straw as mulching material, de Lima et al. (2019) found in a sandy loam soil
597 that an increase in mulch length leads to a decrease in surface cover and then in soil
598 erosion rates.

599 The direct associations among the hydrological variables (runoff and soil loss), mulch
600 parameters and soil cover found in the current study are consistent with numerous
601 results (e.g., Donjadee & Tingsanchali, 2016; Won et al., 2012; Yanosek et al., 2006),
602 which showed that, in soils with lower surface cover (generally with increasing mulch
603 length), erosion expectedly increases.

604

605 *4.2. Modeling runoff volume and soil loss using mulch parameters*

606

607 The current study went further in the evaluation of runoff and soil loss after rainfall
608 simulation under different mulch conditions, proposing prediction models of these
609 hydrological variables. The multiple-regression analysis has indicated that surface
610 runoff and soil loss can be estimated from the mulch parameters using simple but
611 powerful equations with a linear mathematical form. The input data of these models are
612 simply the mulch application rates and lengths. Therefore, for a given precipitation
613 depth and intensity (as that used for these experiments), the models predict both the
614 runoff volume and soil loss. The values of the regression coefficients of the developed
615 equations show that the mulch application rate has much more influence than straw
616 length (the ratio between these parameters is equal to about 200 for runoff and 400 for
617 soil loss) and the interaction factor (that is, the product of mulch application rate by
618 length) has a very low influence on the predicted variables. This result is consistent with
619 the findings of Lal (1976), who demonstrated that the mulch application rate can be
620 assumed as predictor of surface runoff and soil loss, both being significantly ($p < 0.05$)
621 influenced by the mulch parameters. Clearly, the intercepts of the two equations are the
622 runoff and soil loss expected under bare soil conditions. The model coefficients of ML
623 and MAR are negative for runoff, since the latter decreases when the mulch application
624 rate increases. Instead, these coefficients are discordant (negative for MAR and positive
625 for ML) for soil loss, as erosion increases with coarser particles of straw and decreases
626 for higher doses of mulch.

627 The developed equations are related to the precipitation variables (rainfall depth and
628 intensity) that have been used under the simulated rainfall experiments. Therefore, for
629 broader applications of these prediction models, a set of equations must be developed
630 for different precipitation characteristics. For instance, having an intensity-duration-
631 frequency curve, which gives the rainfall depth and intensity with a given return interval
632 (that is, with a desired probability), the values of the regression coefficients can be
633 calibrated. This helps land managers in soil conservation issues, which are pressing
634 particularly in deforested areas, as those of the current study.

635 The developed models could be applied by two approaches. First, the most suitable
636 application rate and length of mulch needed to keep the modeled hydrological variables
637 under a tolerance limit, which, for soil loss, is in the range 3 - 11.2 Mg/ha · yr (Bazzoffi,
638 2009; Wischmeier & Smith, 1978). Setting up, for instance, this tolerance limit, the
639 prediction model gives the application rate and length of rice straw mulch, which have
640 to be applied to the soil. Second, these models can be used in combination with other
641 erosion prediction tools, such as the well-known Universal Soil Loss Equation (USLE,
642 Wischmeier, 1973). For instance, Eq. 3 can be used to evaluate the effect of the soil
643 management (mathematically modeled by the USLE C-factor) on the annual soil loss,
644 using experimental plots with the same geomorphological and climatic characteristics,
645 but different application rates and length of rice straw mulch. The current modeling
646 approach should go further with comparison of different straws (such as oat, barley,
647 wheat) and under different slope and soil conditions.

648 In view of transferring the results of the current study to common soil conservation
649 practice, some issues should be taken into account, such as the upscaling effects of the
650 mulch efficacy when increasing the plot length to the hillslope scale. For instance,
651 higher erosion rates can be observed on longer slopes, due to concentration of overland
652 flow with increased sediment transport capacity (Rahma et al., 2017), while Prats et al.
653 (2016), although working on soils deforested by fire, showed that smaller plots can
654 overestimate runoff and erosion when compared to a hillslope scale. Another important
655 issue that is likely to affect land management using straw mulch may be the risk of
656 mulch failure over long hillslopes due to the removal effect of runoff. This risk could be
657 evaluated by applying a modeling approach helping to identify the maximum length of
658 slope that can be effectively protected by mulch without increased runoff and erosion
659 rates.

660

661 **5. Conclusions**

662

663 Under simulated rainfall on a deforested soil treated with rice straw mulch with
664 different application rates and lengths, runoff and soil loss in mulched soils were
665 significantly ($p < 0.05$) lower than the corresponding variables observed for bare soil.
666 The lowest runoff was observed for a mulch layer of 3 Mg/ha of straw with length of
667 200 mm. The lowest soil loss was found with the same application rates but with 10 mm
668 length. These outcomes confirm one of the working hypotheses that higher application
669 rates of rice straw generate less runoff and soil erosion, but reject, at least for the soil
670 loss, the other hypothesis that to reduce the soil loss the length of rice straw must be
671 long. The multiple-regression equations, developed to predict runoff and erosion as a
672 function of mulch application rate and length, show very good accuracy and can be used
673 as prediction models for identifying the most suitable mulch parameters for effective
674 soil protection.

675

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679

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