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Compost and vermicompost in cucumber rhizosphere promote plant growth and prevent the entry of anthropogenic organic pollutants

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*Original*

Compost and vermicompost in cucumber rhizosphere promote plant growth and prevent the entry of anthropogenic organic pollutants / Carnimeo, C., Gelsomino, A., Cirrottola, G., Panuccio, M.R., Loffredo, E..  
- In: SCIENTIA HORTICULTURAE. - ISSN 0304-4238. - 303:(2022), p. 111250.  
[10.1016/j.scienta.2022.111250]

*Availability:*

This version is available at: <https://hdl.handle.net/20.500.12318/127005> since: 2024-09-26T14:02:00Z

*Published*

DOI: <http://doi.org/10.1016/j.scienta.2022.111250>

The final published version is available online at: <https://www.sciencedirect>.

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(Article begins on next page)

1                    *This is the peer reviewed version of the following article:*

2  
3     ***Parhizkar, M., Shabanpour, M., Lucas-Borja, M. E., Zema, D. A., Li, S., Tanaka, N.,***  
4     ***& Cerda, A. (2021). Effects of length and application rate of rice straw mulch on***  
5     ***surface runoff and soil loss under laboratory simulated rainfall. International***  
6                    ***Journal of Sediment Research, 36(4), 468-478,***

7  
8                    *which has been published in final doi*

9  
10                    10.1016/j.ijsrc.2020.12.002

11  
12  
13                    (<https://www.sciencedirect.com/science/article/pii/S100162792030127X>)

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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  $160 \times 10^6$  ha globally, with most of the  $700 \times 10^6$  t world production grown in  
115 Asia ( $640 \times 10^6$  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**

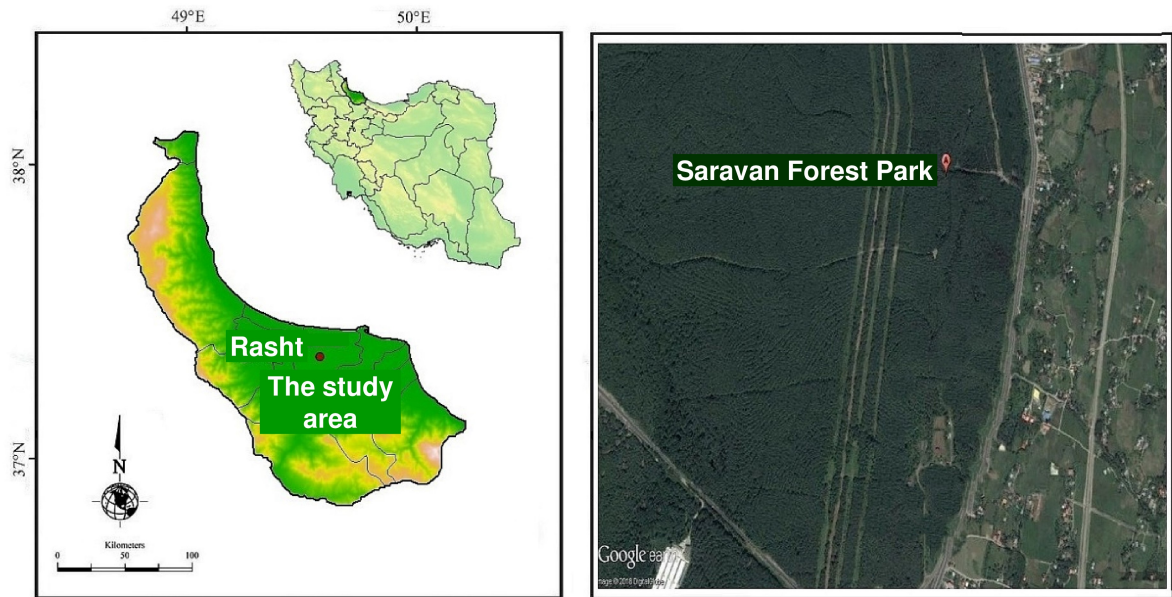
162

### 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<sup>®</sup> Maps<sup>®</sup>) (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  $0.21 \pm 0.03$ . 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  $98 \pm 1.1$  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  $1353 \pm 279$  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/h<sup>1</sup> 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).

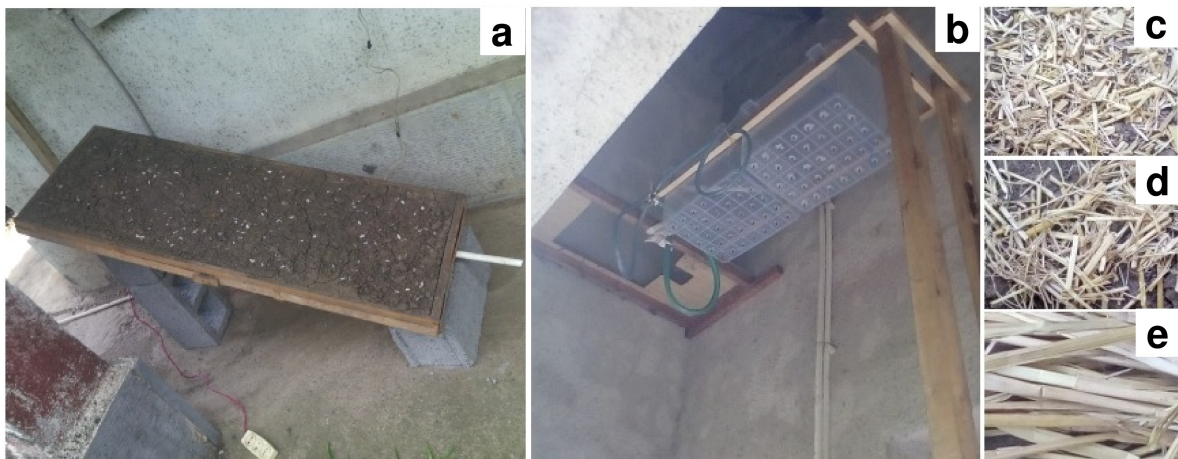
258

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/h<sup>1</sup> 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 ( $13.2 \pm 0.23$  mm), while the  
360 lowest value was observed in the AR3-L200 plots ( $7.62 \pm 0.12$  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 ( $11.31 \pm 0.10$  mm in AR1, to  $8.49 \pm 0.05$  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  $10.67 \pm 0.12$  mm in L10 to  $9.04 \pm 0.06$  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).

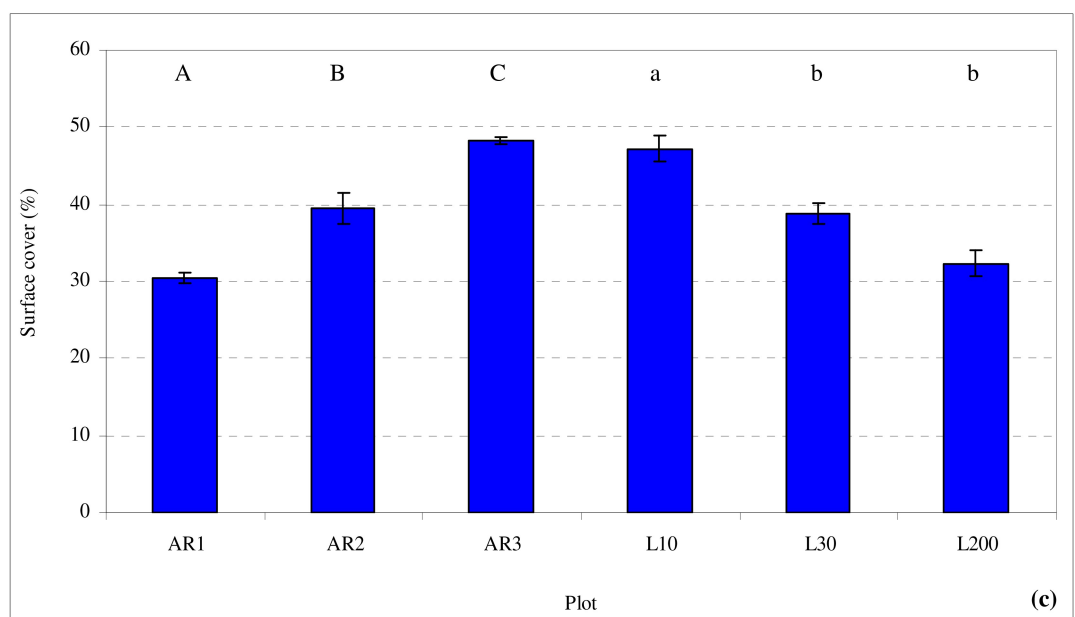
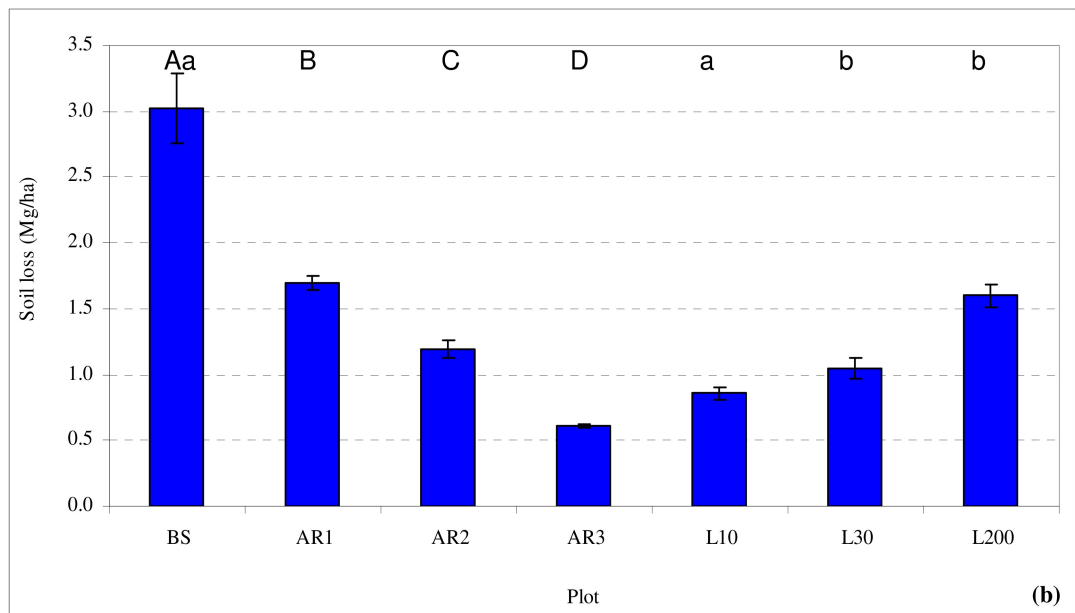
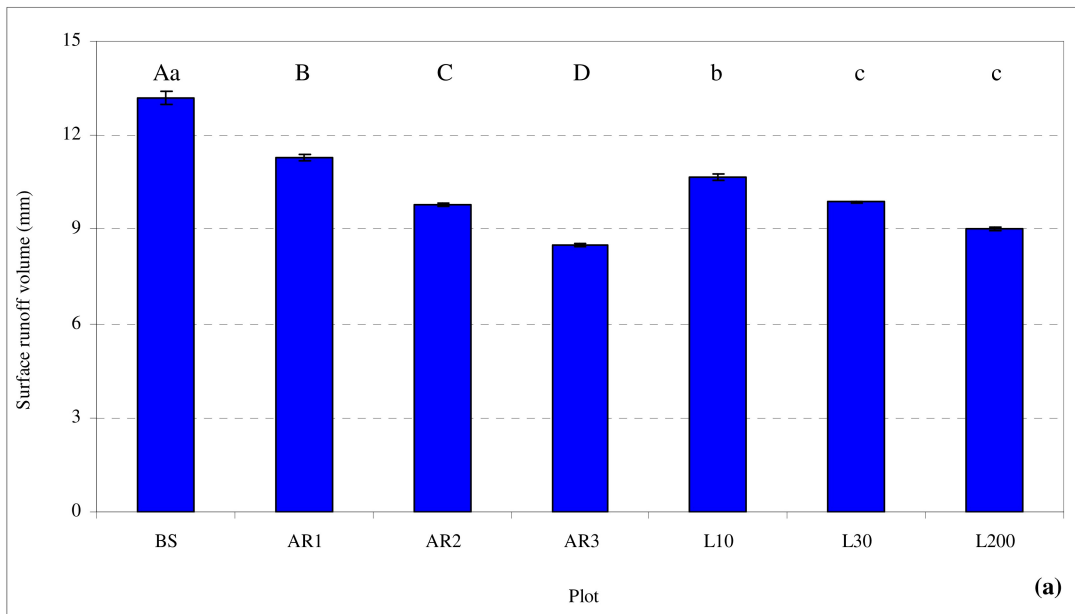
374

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 ( $3.02 \pm 0.2$  Mg/ha). The lowest erosion was  
394 measured in AR3-L10 plots ( $0.42 \pm 0.04$  Mg/ha) (Table 2). It is also interesting to note  
395 that a high soil loss ( $2.28 \pm 0.20$  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  $1.70 \pm 0.05$  Mg/ha in AR1 to  $0.61 \pm 0.01$  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  $1.60 \pm 0.09$  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	<b>0.842</b>	<b>-0.909</b>	<b>-0.872</b>
Mulch length		1	-0.099	<b>-0.516</b>	0.162
Surface cover			1	<b>-0.649</b>	<b>-0.948</b>
Runoff volume				1	<b>0.663</b>
Soil loss					1

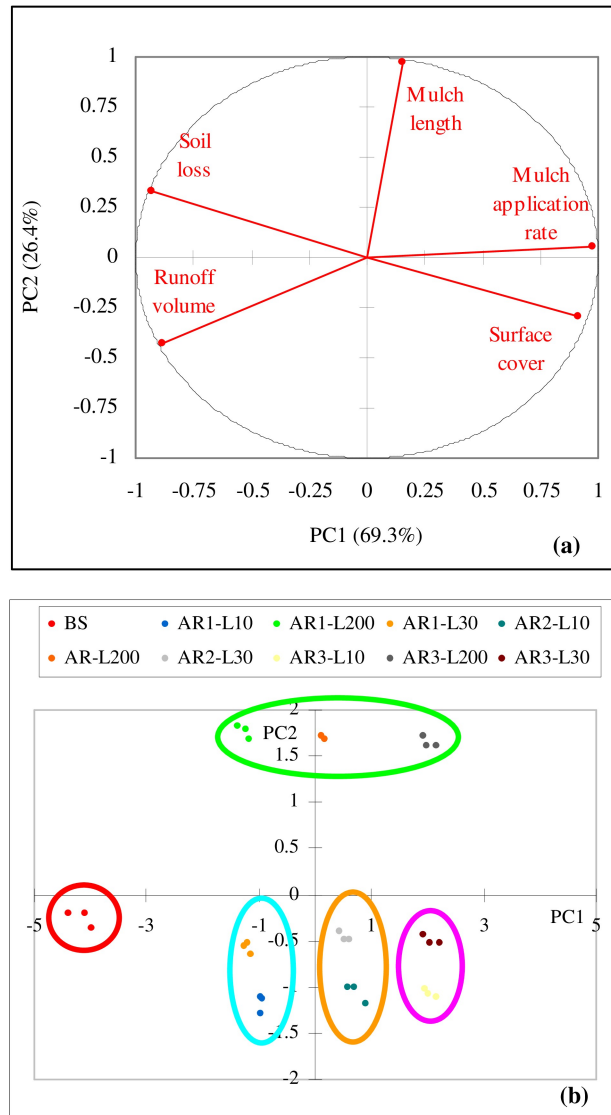
434

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) (PC<sub>1</sub> and PC<sub>2</sub>) (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

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

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

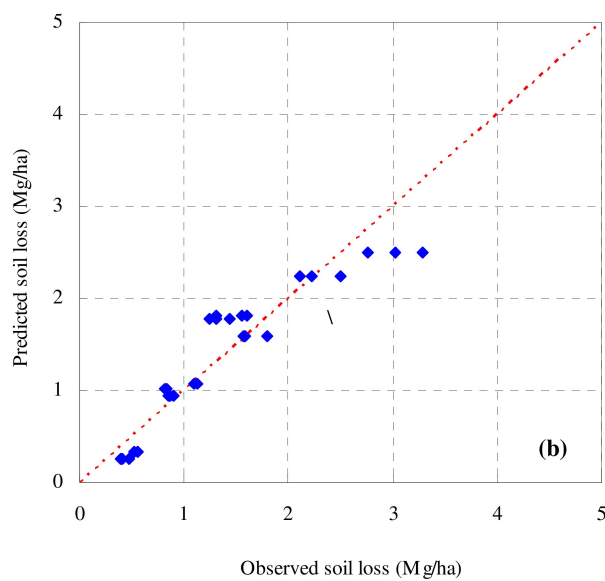
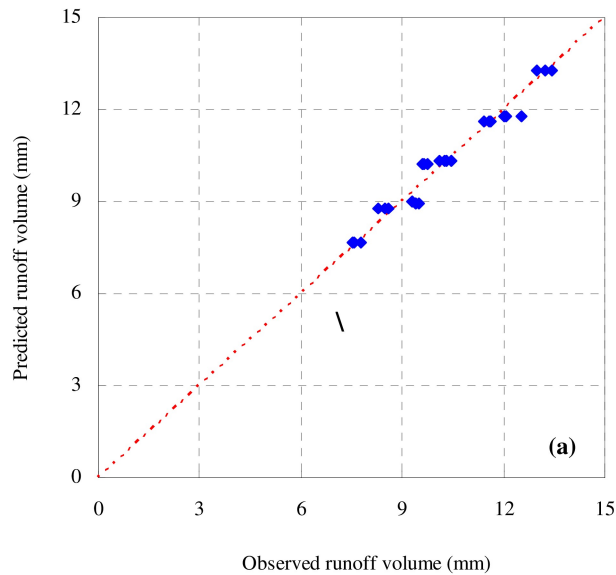
474

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 <sup>2</sup>	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; R<sup>2</sup> = 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

676 **Acknowledgments:** The authors thank the Faculty of Agricultural Sciences, University  
677 of Guilan for their support and experimental assistance and two anonymous Reviewers,  
678 whose suggestions really helped to improve the paper.

679

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681

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