Assessment of crusting effects on interrill erosion by laser scanning

Yaxian Hu [Corresp., 1, 2], **Wolfgang Fister** [2], **Yao He** [1], **Nikolaus J. Kuhn** [2]

[1] State Key Laboratory of Soil Erosion and Dryland Farming on the Loess Plateau, Institute of Soil and Water Conservation, Northwest A&F University, Yangling, Shaanxi, China

[2] Physical Geography and Environmental Change, Department of Environmental Sciences, University of Basel, Basel, Switzerland

Corresponding Author: Yaxian Hu
Email address: huyaxian@nwafu.edu.cn

Background. Crust formation affects soil erosion by raindrop impacted flow through changing particle size and cohesion between particles on the soil surface, as well as surface microtopography. Therefore, changes in soil microtopography can, in theory, be employed as a proxy to reflect the complex and dynamic interactions between crust formation and erosion caused by raindrop-impacted flow. However, it is unclear whether minor variations of soil microtopography can actually be detected with tools mapping the crust surface, often leaving the interpretation of interrill runoff and erosion dynamics qualitative or even speculative.

Methods. In this study, we used a laser scanner to measure the changes of the microtopography of two soils placed under simulated rainfall in experimental flumes and crusting at different rates. The two soils were of the same texture, but under different land management, and thus organic matter content and aggregate stability. To limit the amount of scanning and data analysis in this exploratory study, two transects and four subplots on each experimental flume were scanned with a laser in one-millimeter interval before and after rainfall simulations.

Results. While both soils experienced a flattening, they displayed different temporal patterns of crust development and associated erosional responses. The laser scanning data also allowed to distinguish the different rates of developments of surface features for replicates with extreme erosional responses. The use of the laser data improved the understanding of crusting effects on soil erosional responses, illustrating that even limited laser scanning provides essential information for quantitatively exploring interrill erosion processes.

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7 [1] State Key Laboratory of Soil Erosion and Dryland Farming on the Loess Plateau, Institute of
8 Soil and Water Conservation, Northwest A&F University, Yangling, Shaanxi, China
9 [2] Physical Geography and Environmental Change, Department of Environmental Sciences,
10 University of Basel, Basel, Switzerland
11
12 Corresponding Author:
13 Yaxian Hu[1]
14 Xinong Road 26, Yangling, Shaanxi, 712100, China
15 Email address: huyaxian@nwafu.edu.cn
16

Abstract

17
18 **Background.** Crust formation affects soil erosion by raindrop impacted flow through changing
19 particle size and cohesion between particles on the soil surface, as well as surface
20 microtopography. Therefore, changes in soil microtopography can, in theory, be employed as a
21 proxy to reflect the complex and dynamic interactions between crust formation and erosion
22 caused by raindrop-impacted flow. However, it is unclear whether minor variations of soil
23 microtopography can actually be detected with tools mapping the crust surface, often leaving the
24 interpretation of interrill runoff and erosion dynamics qualitative or even speculative.
25 **Methods.** In this study, we used a laser scanner to measure the changes of the microtopography
26 of two soils placed under simulated rainfall in experimental flumes and crusting at different
27 rates. The two soils were of the same texture, but under different land management, and thus
28 organic matter content and aggregate stability. To limit the amount of scanning and data analysis
29 in this exploratory study, two transects and four subplots on each experimental flume were
30 scanned with a laser in one-millimeter interval before and after rainfall simulations.
31 **Results.** While both soils experienced a flattening, they displayed different temporal patterns of
32 crust development and associated erosional responses. The laser scanning data also allowed to
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35 soil erosional responses, illustrating that even limited laser scanning provides essential
36 information for quantitatively exploring interrill erosion processes.
37

Introduction

39 After decades of improvement by researchers and farmers worldwide, risks of field-scale soil
40 loss can be predicted reasonably well by erosion models such as the Revised Universal Soil Loss
41 Equation (RUSLE) (Wischmeier & Smith, 1978) or the Water Erosion Prediction Project
42 (WEPP) (Flanagan & Nearing, 2000). In the latter model, and several other process-based
43 models (e.g., EUROSEM, CREAMS), the contribution of interrill erosion to the overall soil loss
44 has been accounted for in separate sub-models (Knisel & Nicks, 1980; Morgan et al., 1998;
45 Aksoy & Kavvas, 2005). Interill erosion comprises soil loss by non-concentrated and raindrop-
46 impacted flow (Kinnell, 2005). Fine and light particles enriched in organic matter are often
47 entrained and transported selectively, leading to preferential removal of soil organic carbon and
48 phosphorus, and their transfer into watercourses as non-point source pollution (Sharpley, 1985;
49 Quinton, Catt & Hess, 2001; Lal, 2003; Teixeira & Misra, 2005; Warrington et al., 2009; Kuhn
50 et al., 2012; Hu, Fister & Kuhn, 2013). Therefore, realistic simulation of interrill erosion is
51 essential. A key problem for such modeling is the parameterization of soil resistance to crust
52 formation during one or a sequence of erosion events (Kuhn, Bryan & Navar, 2003; Kuhn &
53 Bryan, 2004; Kinnell, 2005; Hu, Fister & Kuhn, 2016).
54 When subjected to raindrop impact, the soil surface is compacted, and soil aggregates experience
55 destruction through slaking, swelling, micro-cracking and dispersion (Le Bissonnais, Bruand &
56 Jamagne, 1989; Le Bissonnais, 1996; Darboux & Le Bissonnais, 2007). The overall impacts of

57 these changes of soil surface properties on interrill erosion are ambiguous because they can
58 induce the formation of cohesive structural crusts, which stabilize the soil surface and protect it
59 from further erosion (Chen et al., 1980; Le Bissonnais, 1990). However, a compacted surface is
60 also smoother, leading to shallower and faster runoff, which increases runoff erosivity (Le
61 Bissonnais, Renaux & Delouche, 1995; Quinton, Catt & Hess, 2001; Kuhn, 2007). Meanwhile,
62 the destruction of aggregates generates fine, loose mineral or aggregated fragments that can be
63 transported easily by raindrop-impacted thin flows (Kuhn, Bryan & Navar, 2003; Anderson &
64 Kuhn, 2008). Smoothing of the surface, as well as the formation of local surface irregularities
65 also induce differences in runoff depth and thus the way raindrop energy is dissipated on soil
66 surface (Kinnell, 2005). Depending on the flow depth, this may protect soils from erosion, but
67 can also enhance detachment and transport (Torri, Sfalanga & Chisci, 1987).
68 The effect of crust formation is not limited to the amount of erosion, but also affects the quality
69 of sediment. Due to its limited runoff energy, erosion by non-concentrated flow often moves fine
70 and light particles, including the substances attached to them, selectively (Basic et al., 2002;
71 Schiettecatte et al., 2008; Kinnell, 2012). Aggregates at the soil surface continue to breakdown
72 into smaller fragments during a rainfall event, discharging loose materials of different sizes and
73 densities (Chen et al., 1980). Sediment of varying size compositions is thus enriched differently
74 with nutrients and organic matter, posing unknown pollution risks to downstream watercourses
75 (Kuhn et al., 2012; Hu, Fister & Kuhn, 2013). The variations of sediment composition do not
76 only depend on the changes of stability and transportability of surface materials over time, but
77 are also affected by surface roughness and its interaction with rainfall and runoff (Moore &
78 Singer, 1990; Kuhn, Bryan & Navar, 2003). Furthermore, on soils with mixed texture and stable
79 small aggregates, the initial removal of loose minerals and aggregated particles progressively
80 exposes cohesive structural crust, leading to a distinct temporal pattern of initially high and then
81 declining erosion rates (Moore & Singer, 1990; Kuhn, Bryan & Navar, 2003). In addition, as
82 repeatedly reported in previous studies (Nearing, Govers & Norton, 1999; Armstrong et al.,
83 2011; Hu, Fister & Kuhn, 2016), the inter-replicate variability of runoff and soil erosion rates
84 under strictly controlled experimental conditions also illustrates a complex, fine-scale interaction
85 between soil surface microtopography, crust formation and soil particles, including their spatial
86 patterns, that is hardly overcome even by adding more replicates.
87 The above-described examples of the interaction between crust formation and interrill erosion
88 illustrate the need for spatial data capturing the resistance to erosion (hereafter referred to as
89 erodibility) of soil surface and its change over time. Currently, the lack of such quantitative
90 information introduces uncertainties for interrill erosion modeling (Darboux & Le Bissonnais,
91 2007; Bremenfeld, Fiener & Govers, 2013; Hu, Fister & Kuhn, 2016). Apart from characterizing
92 the composition and stability of the crust, the change of soil surface roughness over time is an
93 essential variable controlling soil resistance to interrill erosion and sediment quality (Kuhn,
94 Bryan & Navar, 2003; Armstrong et al., 2012; Kuhn & Armstrong, 2012). However, soil
95 microtopography changes are difficult to monitor (Issa et al., 2004; Algayer et al., 2014). Studies
96 of soil micromorphology, e.g. Chen et al. (1980), do not capture the appropriate scale, whilst

97   indirect measurements of roughness on larger areas, for example by detecting directional
98   reflectance of crusts (Anderson & Kuhn, 2008; Croft, Anderson & Kuhn, 2012), only provide an
99   index value for roughness. Other recently developed imaging techniques, such as Structure from
100 Motion (SfM), have also been applied to detect small soil surface variations (Vinci et al., 2017;
101 Krenz & Kuhn, 2019) or to point out the spatial distribution of erosion and deposition hotspots at
102 plot scale (Remke et al., 2016; Krenz & Kuhn, 2019). Terrestrial or airborne laser scanning have
103 also been widely used to investigate slope-scale rill erosion or catchment-scale gully morphology
104 (Vinci et al., 2015; Wu et al., 2017). However, all these studies and methods lack the millimeter-
105 scale resolution required for the assessment of crust formation and its effects on flow hydraulics
106 and soil resistance to erosion (Eltner et al., 2015). A method such as laser scanning, at least in a
107 laboratory setting, enables the generation of the most precise digital elevation models. Huang &
108 Bradford (1992) demonstrated that laser scanning at a 0.5-mm grid resolution can be used to
109 replace speculative explanations of soil erosional responses with more quantitative
110 interpretations. Abban et al. (2017) employed a laser scanner and several indices derived from
111 the obtained data to decipher the role of rain splash on surface roughness. All these studies
112 illustrate that laser scanning has the potential for closing the gap between the ideal quality of data
113 required to assess crusting effects on interrill erosion and a feasible acquisition of soil
114 microtopography in a laboratory setting.
115 In this study, we carried out high-resolution laser scanning on a laboratory flume to measure
116 changes in crust microtopography before and after a rainfall event. The data were obtained
117 during an experiment that focused on the temporal dynamics of soil erosion by raindrop-
118 impacted flow on two soils of similar texture, but with different land management, organic
119 matter content and aggregate stability (Hu, Fister & Kuhn, 2013). We hypothesized that the high-
120 resolution laser scanning signals were sufficiently sensitive to detect minor changes of soil
121 microtopography induced by crust formation, and thus can be employed to quantitatively explain
122 differences of soil erosional responses between soils subject to different land management. Such
123 use of laser scanning would avoid the often qualitative explanations for differences in interrill
124 erosion between events or soils of different quality.
125

Materials & Methods

127 **Experimental design and soil sampling**
128 Two silty loams of similar texture, but different soil organic carbon content (SOC) and aggregate
129 stability (Table 1), were subject to prolonged rainfall simulations in this study. The surface of the
130 two soils were laser scanned before and after the rainfall events to detect the crusting-induced
131 changes of soil microtopography and the effect on erosion by raindrop-impacted flow. As part of
132 a series of experiments, detailed information on soil properties and experimental design have
133 been described in Hu et al. (2013). In brief, the two soils, one conventionally (CS) and one
134 organically managed (OS), were sampled in 2010 from A-horizons on two farms near Möhlin
135 (47°33'N, 7°50'E) in northwest Switzerland. After drying at 40°C, the two soils were sieved into
136 1 to 8 mm to exclude over-sized clods and to allow the observation of relative changes in surface

137   roughness during erosion processes. Immediately after sieving, each of the soils was filled into a
138   round flume (Figure 1a), with an outer ring of 50 cm in diameter and an opening of 10 cm in the
139   center (Figure 1b). The slope between the outer edge and the inner opening was 10%. To
140   facilitate drainage, the floor of the flumes was perforated and covered by a fine cloth and a layer
141   of sand (approximately 2 cm). This design of the flumes created a relatively uniform pattern of
142   erosion processes along a short slope with a surface area that generated sufficient runoff and
143   sediment for further analysis.
144
145   **Prewetting and prolonged rainfall simulations**
146   In order to facilitate the generation of runoff and enable the observation of the effects of
147   aggregate breakdown during crusting over the actual rainfall events on the following day, the
148   two flumes were first subjected to a pre-wetting rainfall at an intensity of 30 mm h^{-1} for 30 min
149   right after the soil had been placed into the flume (Figure 1c). The intensity of the actual rainfall
150   was also 30 mm h^{-1}, but it lasted for 360 min, long enough to achieve prolonged steady-state
151   runoff and associated erosional responses. A Fulljet nozzle (Spraying Systems 1/4 HH14WSQ)
152   was used to generate multiple-sized raindrops (D$_{50}$ of 2.3 mm) with an average kinetic energy of
153   113.9 J m^{-2} h^{-1} (measured by a Joss Waldvogel-Disdrometer). While the lower kinetic energy of
154   the rainfall compensates to a certain extent for the high intensity, the design of the experiment
155   did not aim at the strict simulation of natural conditions, but focused on identifying the effects of
156   crust formation on erosion and sediment properties over time.
157   During the 360 min rainfall events, runoff and sediment samples were collected every 30 min
158   from the inner ring in the center. The total weight of the sediment suspension and the dry weight
159   of the sediment particles were measured to calculate runoff and soil erosion rates. The SOC
160   concentration of each sediment sample was determined using a LECO RC 612 at 550°C. The
161   enrichment ratio of eroded SOC (ERsoc) was calculated between the SOC concentration of the
162   eroded sediment and that of the original soil (Hu, Fister & Kuhn, 2013). After the 360 min
163   prolonged rainfall events, the round flumes were oven-dried at 40°C, and the loose material on
164   each round flume was then collected by a vacuum cleaner and weighed (after scanning). The
165   entire experiment was repeated twelve times, but due to technical failure, only the data from the
166   latter 10 replicates (replicate number 3 to 12) were used in this study.
167

168   **Laser scanning of soil surface**
169   To capture the changes of soil surface induced by rainfall and erosion, each flume was laser-
170   scanned three times (Figure 1d): first in the dry condition after filling the soil in the flume
171   (hereafter termed as "Before"); second, after the 30 min prewetting event (hereafter termed as
172   "Prewetted"); and finally, after the 360 min rainfall event (hereafter termed as "After"). The laser
173   was connected to a laptop and data recording was carried out using a MatLab routine. The
174   scanned data were recorded as x-, y-, and z-values.
175   During each laser scanning, two transects (a, b) and four sub-plots (a, b, c, d, each 5 cm × 18 cm)
176   were scanned (Figure 1e). The two transects crossed at the center of the flume, and each transect

177 was scanned stepwise at a 1 mm resolution, in sequence detecting 500 scanning points (Figure
178 1e). On each of the four subplots (each 90 cm^2), the 500 scanning points were visited
179 automatically following the path calculated by the scanning system, and in total covered about
180 19% of the entire eroding area (1884 cm^2) (Figure 1e). Since similar raindrop-impacted flow
181 would be dominant on the erosion plots, the soil surface microtopography was presumed to be
182 spatially uniform on each plot. Therefore, two transects and four subplots were considered
183 adequately representative to investigate erosion-induced changes of surface microtopography. In
184 order to ensure precise alignment, the flume was accurately positioned at bottom left-hand corner
185 of the scanning frame prior to each scanning run (Figure 1d) and accordingly coordinated as (0,
186 0). By doing so, we made sure that each time the same area of the flume was scanned.
187 The laser scanner used in this study (Figure 1d) was custom-designed and built at the University
188 of Basel based on a design used by Anderson & Kuhn (2008) and Croft, Anderson & Kuhn
189 (2009). It consisted of a combined laboratory laser and sensor, manufactured by Baumer Electric,
190 Frauenfeld, Switzerland (model number OADM 2014471/S14C), working at a wavelength of
191 675 nm and suited to measure distances of up to five meters at an accuracy of 0.1 mm (Brunton,
192 2004). The laser was mounted in one-by-one metre frame where it could be moved with stepper
193 motors in a pre-programmed way using a Stepper Motor Controller CSD 315 (Isel Automation,
194 Germany).
195

196 **Calculation of soil surface elevation across two transects and four subplots**
197 Since the laser scanning was conducted stepwise over two transects and four subplots with strict
198 alignment, the linear route of the former and the programmed route of the latter enabled a
199 pairwise comparison of the elevation of individual scanning points. In addition, although interrill
200 erosion did not form concentrated flow on the small round flume used in this study, the tilted soil
201 surface provided a predominant runoff direction, which was parallel to the scanned transects.
202 Therefore, with limited scanning data and pairwise comparison, the height distribution of
203 scanning points was considered adequate to serve the purpose of comparing the minor variations
204 in surface microtopography induced by crusting and their impacts on soil erosional responses
205 (Croft, Anderson & Kuhn, 2009; Vinci et al., 2015).
206 By conducting regression analysis with the laser signals (the space between the scanner frame
207 and the soil surface), the actual height of each point was calculated using Eq. (1):
208 $H_i = max\{z_1, z_2, z_3, \cdots, z_{500}\} - (z_i \times 4.8619 + 0.1491)$ $\qquad$ (1)
209 Where, H_i is the height of each data point (mm); z_i is the distance between the laser and soil
210 surface of each data point; $max\{z_1, z_2, z_3, \cdots, z_{500}\}$ is the longest distance between the laser and
211 soil surface of all the points (namely the zero-level elevation); i is in sequence from the 1st to the
212 500th scanning point; the constants 4.8619 and 0.1491 are the regression coefficients.
213 To exclude the distorted points around the opening in the center, as well these immediately
214 nearby the outer ring (Figure 1b), only two subsections on each side of the two transects were
215 analyzed: $91 \le i \le 200$ on the left half and $311 \le i \le 420$ on the right half of the X axis.

216   Furthermore, to eliminate the bias introduced by the original slope steepness and to better reflect
217   the relative surface height changes at local scale, the heights of each data point along the two
218   transects were standardized by the slope steepness and the distance from the lowest edge to the
219   targeted point (Eq. 2, 3). The sample protocol was also applied for the subsections of the upper
220   and lower half of the Y axis.

221   $h_i = H_i - \left(\max\{x_{91}, x_{92}, x_{93}, \cdots, x_{200}\} - x_i \right) \times 10\%$ (2)

222   or,

223   $h_i = H_i - \left(x_i - \min\{x_{311}, x_{312}, x_{313}, \cdots, x_{420}\} \right) \times 10\%$ (3)

224   As there were no center opening or edge effects on the four subplots (Figure 1e), all the 500 data
225   points of each subplot were analyzed. Since the bias possibly introduced by the original slope
226   steepness of 10% was systematic and limited to the four subplots with small areas (5 cm $\times$ 18
227   cm), the H_i of each scanning point inside the subplots was not standardized to slope steepness in
228   this study. Moreover, to quantitatively compare height distributions in the four subplots, all the
229   measured heights were then classified into eight height classes: < 3 mm, 3-4 mm, 4-5 mm, 5-6
230   mm, 6-7 mm, 7-8 mm, 8-9 mm and > 9 mm. To visualize the changes of surface roughness after
231   erosion events, the variogram analysis of the four subplots were conducted using GS+
232   (Geostatistics for the Environmental Sciences). Kriging regression was applied to give the best
233   linear unbiased prediction of the intermediate values, which were then employed to plot a 2-D
234   version of soil surface height distribution for each subplot. In addition, the height differences
235   between Before, Prewetted and After tests with the least and most eroded replicates were also
236   compared to detect whether the erosion processes were the same, but just operating at different
237   rates, or whether the soil surfaces developed in different ways and thus leading to different
238   erosion processes.

239

240   **Results**
241   **Soil erosional responses and enrichment ratio of eroded SOC**
242   The soil erosional responses observed during the experiments are reported in detail in Hu et al.
243   (2013) and only summarized here briefly (Table 2). Rates of runoff and soil erosion, and
244   enrichment ratio of organic carbon in the eroded sediment (ERsoc) showed clear temporal
245   patterns. The runoff on the CS increased and reached a steady state of 12.9 mm h^{-1} ~~state~~ after 180
246   min, while the runoff on the OS required 240 min to stabilize at 10.7 mm h^{-1}. As runoff increased
247   over rainfall time, the soil erosion rate of the CS increased first, peaked when runoff rate reached
248   steady state and decreased afterwards. The ERsoc of the CS peaked at 1.94 around 150 min,
249   while that of the OS reached only 1.44 after 330 min (Table 2). Further information on soil
250   erosional response on the CS and OS are listed in Table 2.
251   A further noteworthy result is the inter-replicate variability, which remained between 15 and
252   39% even after the maximum runoff and erosion were reached (Hu, Fister & Kuhn, 2016). Out
253   of the ten times repeated simulations, the least and most eroded replicate for the CS were CS-4
254   and CS-11, and replicate OS-9 and OS-12 for the OS (Table 3). Typically, the total runoff, soil

255 erosion and SOC loss of the most eroded replicates nearly doubled that on the least eroded
256 replicate, even though they received comparable rainfall amount (Table 3).
257

Changes of soil surface elevations across the two transects

259 Figure 2 shows the changes of the soil surface at different conditions (Before, Prewetted, After
260 and Post-dried). After 360 min prolonged rainfall, the CS surface was visibly smoother with
261 extended flat areas and few loose material (10.96 ± 3.01 g m^{-2} as listed in Table 3), whereas the
262 OS surface was covered to a greater extent by degraded aggregates (43.78 ± 11.40 g m^{-2} as listed
263 in Table 3). The surface elevation changes of the CS and OS are illustrated in Figure 3. For both
264 soils, the soil surface was lowered after the two rainfall events, with a reduction most evident
265 after the 360 min prolonged rainfall (Figure 3).
266 Apart from the average changes of surface elevation, Figure 4 further compares the height
267 differences between the least and most eroded replicate under the three conditions (Before,
268 Prewetted and After). The height differences between the Prewetted and Before were more
269 closely clustered than that between the After and Before (Figure 4a vs. 4b, 4c vs. 4d).
270 Specifically, for the height differences between After and Before on the CS (Figure 4b), the most
271 eroded replicate CS-11 had more negative height differences (point clouds < 0) than increases
272 (point clouds > 0) when compared to the least eroded replicate CS-4. Similar, but more frequent
273 negative height differences, were observed on the most eroded replicate OS-12 than that of the
274 least eroded replicate OS-9. Moreover, the point clouds were also more concentrated under the
275 1:1 ratio for both the pair-comparison of CS-4 against CS-11, and that of OS-9 against OS-12
276 (Figure 4b, 4d).
277

Changes of soil surface elevations on the four subplots

279 Figure 5 shows that the height distribution of all the four subplots on the CS was greater and
280 more variable than that on the OS. While the surface of both soils progressively approached the
281 lower height classes over time (Figure 5), this transition was much more skewed on the CS,
282 especially after the 360 min prolonged rain (Figure 5d). The differences of surface elevation
283 between the two soils are also illustrated by the 2D classification in Figure 6. The surface
284 elevation of the two soils was quite similar when the soils were dry before the rainfall. After the
285 30 min of prewetting, the surface of the OS still showed a strong contrast of high and low
286 elevation (Figure 6d, 6e), whilst the surface of the CS was noticeably flattened (Figure 6a, 6b).
287 After the 360 min prolonged rain, the surface height of the CS became lowered to less than 4 mm
288 (Figure 6c), that of the OS remained rougher between 4.5- and 6.5-mm (Figure 6f).
289

Discussion

Temporal patterns of crusting and erosion on differently structured silty loams

292 Given the limited runoff depth and erosion capacity of thin flow, interrill erosion is mostly
293 attributed to raindrops impacting the flow (Kinnell, 2005). Hu, Fister & Kuhn (2013) speculated

294   that the differences in erosion and sediment properties observed between the two soils tested in
295   this study were associated with crust formation. They hypothesized that the different erosional
296   responses of the similarly textured CS and OS (Table 2) reflect the influence of aggregate
297   stability (Table 1) on surface crusting and in turn, the capacity of raindrop impacted flow for
298   erosion. In particular, testing the hypotheses of Hu, Fister & Kuhn (2013) requires a quantitative
299   assessment of the potential effects of crusting on soil surface microtopographic changes over
300   time. The noticeably more pronounced flattening on the CS and the delayed deformation on the
301   OS (Figure 2, 3, 5) confirm the explanation brought forward by Hu, Fister & Kuhn (2013):
302   greater aggregate stability of the OS slowed aggregate breakdown, maintaining roughness and
303   sediment size for longer, thus also resisting raindrop impact for longer than on the CS. Such
304   potential effects of aggregate stability on surface deformation are reflected by the more
305   pronounced skewing toward the smaller height classes on the CS surface than on the OS surface
306   (Figure 5). Consequently, after the prolonged 360 min rainfall, the CS surface height was
307   noticeably flattened to be less than 4 mm (Figure 5d, 6c), whereas the soil surface of the OS was
308   much rougher between 4.5 mm and 6.5 mm (Figure 6f) and still interspersed by more loose
309   material (Figure 2g, 2h, Table 3). The declining soil erosion rates on the CS after its runoff rate
310   exceeded 12.9 mm h^{-1} (Table 2) indicate that runoff had overcome transport limitation and
311   reached a supply-limited process after fine, light and loose particles had been selectively eroded
312   (flattened surface in Figure 5d, 6c). Judging from the abundant loose materials remaining on the
313   OS plots (Figure 2g, 2h, Table 2, as well in Hu, Fister & Kuhn (2013)), it would also eventually
314   reach a supply-limited condition as runoff grew more competent over time by removing loose
315   particles and exposing cohesive crust. This deduction is also supported by the delayed decline of
316   the ERsoc on the OS once runoff rates had stabilized (Table 2).
317

318   **Variability of crust formation and erosion identified by laser scanning**
319   Apart from detecting the different erosional responses between the two soil types, the laser
320   scanning also effectively captured the variations of surface microtopography among replicates.
321   The greater amount of points with height differences < 0 mm for the most eroded replicate CS-11
322   and OS-12 (Figure 4d) clearly illustrates more advanced smoothening than that for the least
323   replicates CS-4 and OS-9 (Figure 4b). The unbalanced distributions of height differences in the
324   pairwise comparisons (Figure 4) further demonstrate the divergent influences of surface
325   microtopography on the soil erosional responses of the least and most eroded replicates. To be
326   specific, the more concentrated point clouds under the 1:1 ratio line in Figure 4b and 4d
327   practically represent three scenarios. (1) The positive height differences under the 1:1 ratio line
328   in section I of Figure 4b indicate that the scanned surface was rougher after 360 min of rainfall
329   (After) than before, and such roughening was more pronounced in the least eroded replicate CS-
330   4 than the most eroded replicate CS-11. (2) The concentrated cloud in section II of Figure 4b
331   displays that certain parts of the least eroded replicate CS-4 became rougher after the prolonged
332   rainfall than Before (positive height differences from 0 mm to 6 mm), whereas some sections of
333   the most eroded replicate CS-11 were flattened after the prolonged rainfall events (negative

334 height differences from -6 mm to 0 mm). A similar, but even more obvious concentration of the
335 point cloud can be observed in section II of Figure 4d, illustrating the divergent development of
336 soil surface elevation between the least and most eroded replicates of the OS. (3) The negative
337 height differences in section III under the 1:1 ratio line of Figure 4b suggest that those areas were
338 smoother after the 360 min rainfall, and the most eroded CS-11 was more smoothened than the
339 least eroded CS-4. All the three scenarios of more pronounced flattening (Figure 4b, 4d) on the
340 most eroded replicates CS-11 and OS-12 are consistent with their nearly doubled runoff and soil
341 loss as opposed to the least eroded replicates CS-4 and OS-9 (Table 3).
342 The covariance between the observed erosional response and the laser data illustrates the
343 effectiveness of millimeter-resolution laser scanning to detect the minor topographic changes. On
344 the one hand, this not only confirms the decisive role of soil properties such as organic matter
345 content and aggregate stability in crust formation, soil erosional responses and sediment
346 properties (Table 1, 2) (Hu, Fister & Kuhn, 2013). On the other, the results clearly illustrate that
347 laser data also help to uncover the causes of the 15%-39% inter-replicate variability among the
348 ten simulated rainfall events (Hu, Fister & Kuhn, 2016) by effectively distinguishing the rates of
349 surface feature development between the most and least eroded replicates (Figure 4, Table 3).
350 Therefore, with limited laser scanning data and pairwise comparison of height distributions, our
351 observations corroborate previous studies where the changes of soil surface microtopography
352 during rainfall events were effectively quantified by laser scanning (Huang & Bradford, 1992;
353 Abban et al., 2017). Furthermore, our findings show that with laser scanning an improved
354 quantitative interpretation of interrill erosion experiments is possible, which would otherwise
355 have been attributed just in a qualitative way to inherent variability under controlled laboratory
356 conditions (Bryan & Luk, 1981; Anderson & Kuhn, 2008; Hu, Fister & Kuhn, 2016).
357

Conclusions

359 The changes of microtopography of the surface of two soils with different rainfall-erosion
360 interaction were measured by fine-scale laser scanning before and after the application of
361 simulated rainfall. The surface of both soils experienced a flattening, but they displayed
362 persistently different temporal patterns of crust development and associated erosional responses.
363 By effectively distinguishing the minor variations of surface microtopography and thus the rates
364 of surface feature development among the least and most eroded replicates, the height
365 differences detected by laser scanning revealed the causes of the 15%-39% inter-replicate
366 variability among the ten simulated rainfall events. This improved the understanding of crusting
367 effects on soil erosional responses and demonstrated that laser scanning can be used to examine
368 interrill erosion with more quantitative interpretations. While a promising tool in the lab, laser
369 data of the degree of accuracy achieved in this study are likely not readily acquirable under field
370 conditions, especially if plots are large and natural rainfalls determine crust formation (Nearing,
371 1998; Armstrong et al., 2011). However, studies using high resolution laser scanning conducted
372 in the laboratory could be used to develop soil microtopography parameters that can be linked to
373 reflectance data or DEMs derived by Structure from Motion in both laboratory and field and

374   their relationship to interrill erosion (Anderson & Kuhn, 2008; Croft, Anderson & Kuhn, 2009).
375   With the currently rapid development of easily accessible and affordable devices such as smart
376   phones, high-resolution cameras and unmanned aerial vehicles, the associated digital
377   photogrammetry and high-resolution digital elevation models (Eltner et al., 2015; Vinci et al.,
378   2015, 2017), there is a great potential for the development of sensible interrill microtopography
379   parameters and the acquisition of data in laboratory and field to improve soil erosion models.
380
381

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512

Figure 1

Soil materials and experimental design.

(a) Layout of round flumes; (b) cross-section profile of the round flume; (c) Set-up of rainfall simulation experiment; (d) set-up of laser scanner; (e) scanning routes of two transects and four sub-plots.
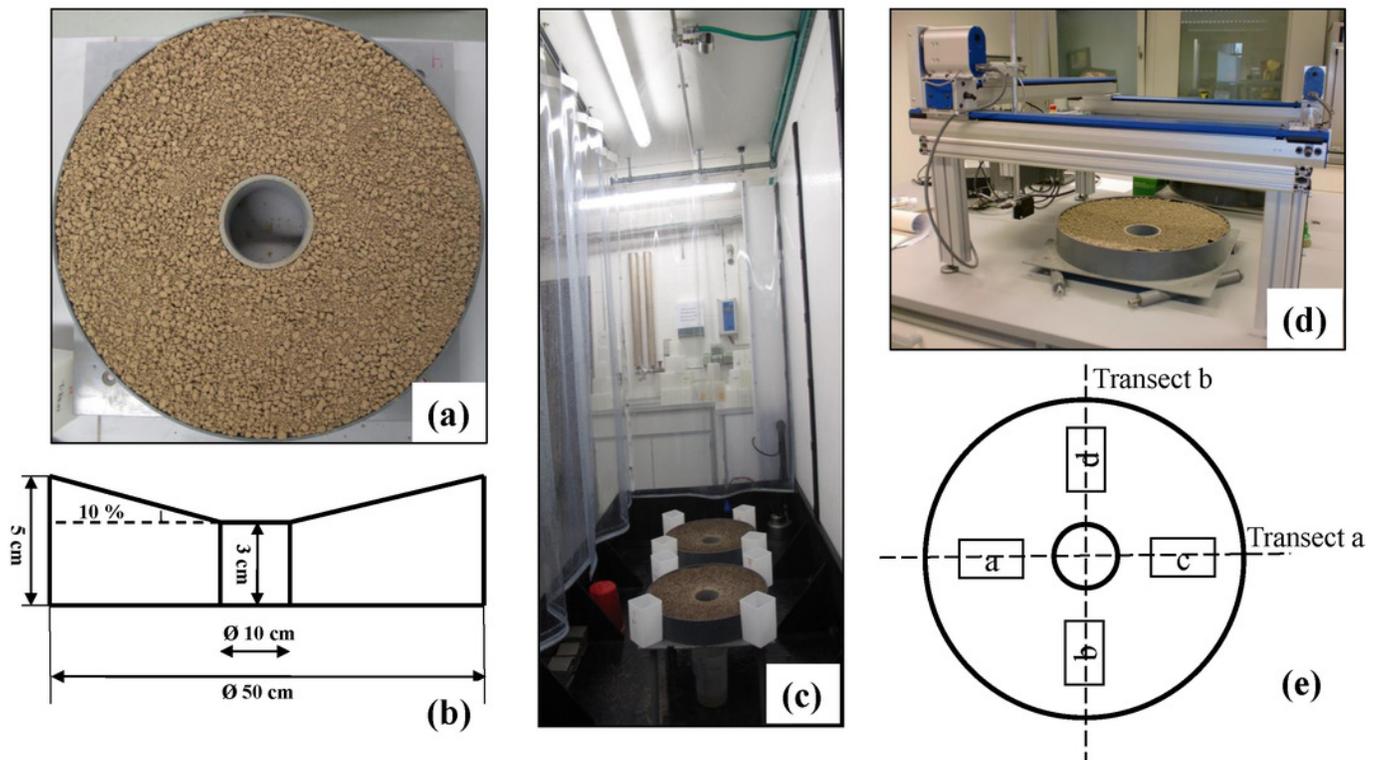
Figure 2

The changes of the two soils (CS and OS) before (Before), prewetted (Prewetted), after the prolonged rainfall events (After), and Posted-dried.

The dark patches are degraded crumbs and blunted coarse aggregates. Light-colored areas are depositional crusts consisting of loose materials detached by raindrop impact.
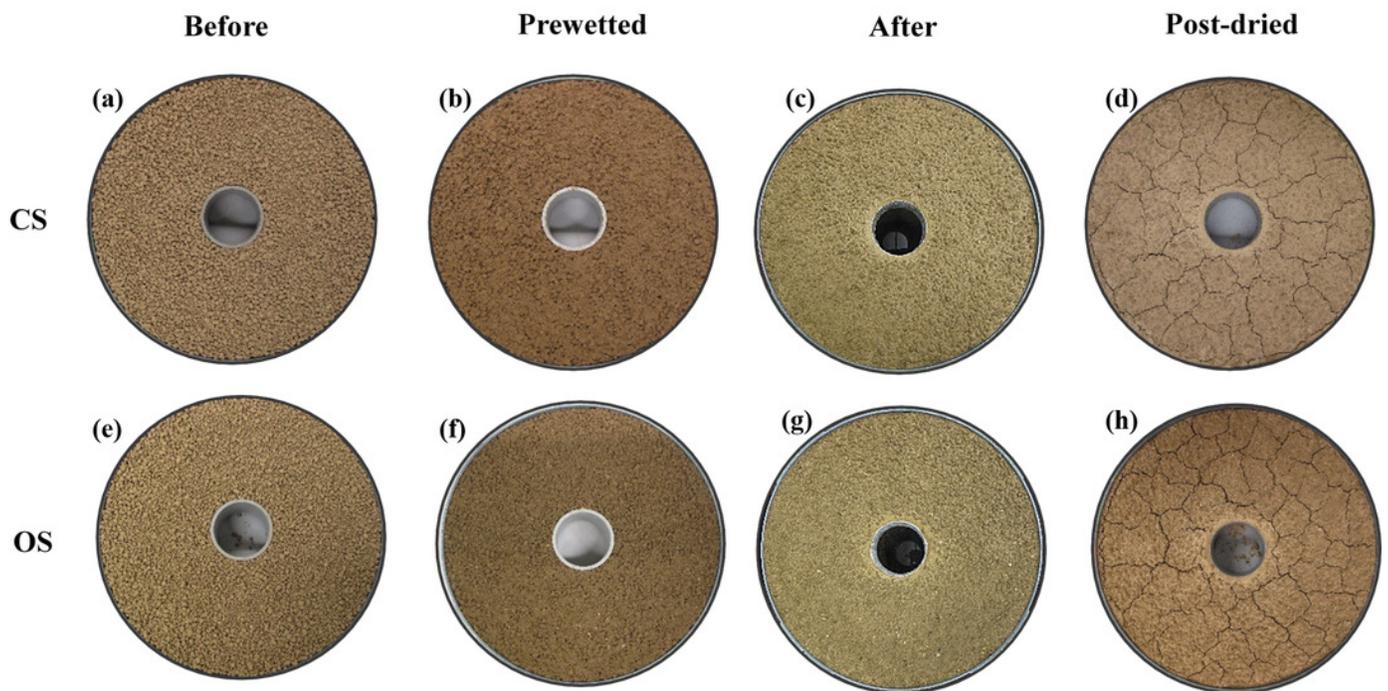
Figure 3

Height of the individual points of the two transects (a, b) of the conventionally farmed soil (CS) and organically farmed soil (OS) Before, Prewetted and After (taking replicate 5 as an example).
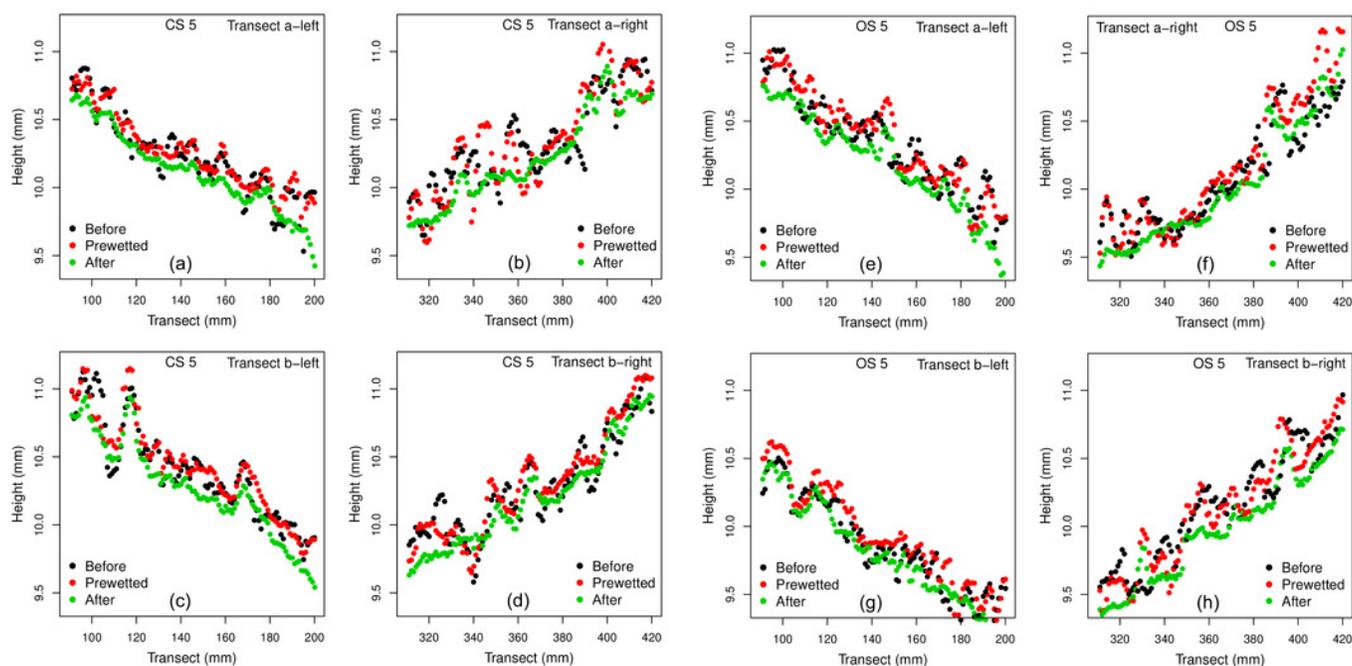
Figure 4

The pair-wise comparison of the normalized height of the two transects on CS-4 vs. CS-11, and OS-9 vs. OS-12.

The replicate CS-4 and OS-9 generated the least soil loss out of the then replicates, whereas the replicate CS-11 and OS-12 produced the most (more information in Table 3). The diagonal line in each subfigure represents the 1:1 ratio.
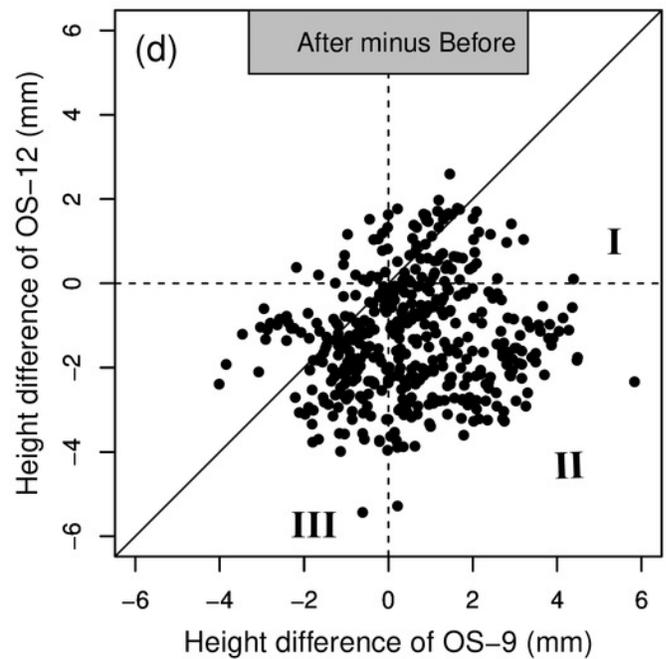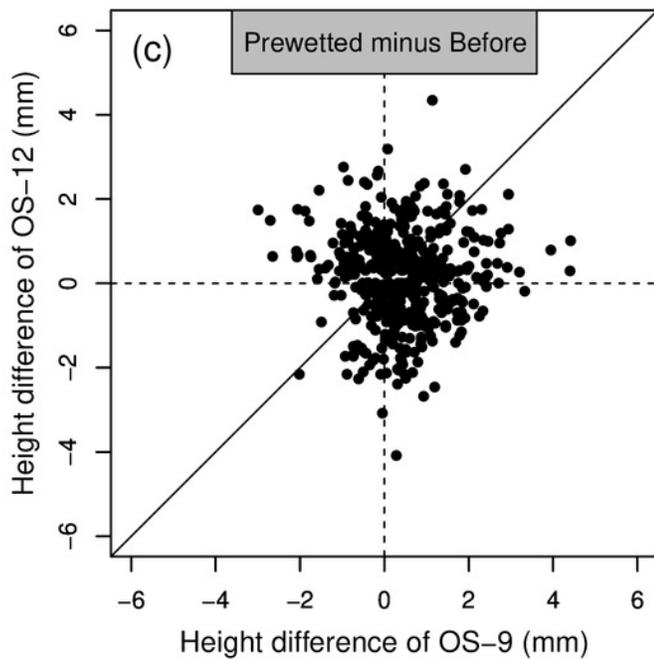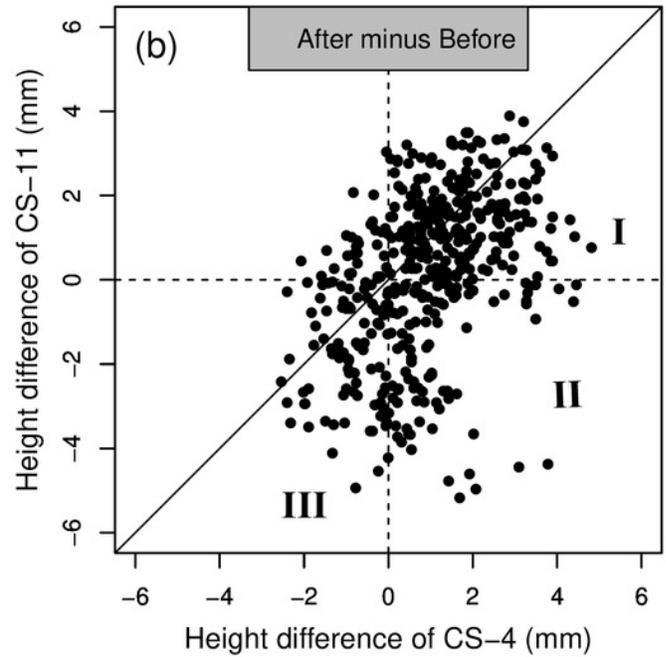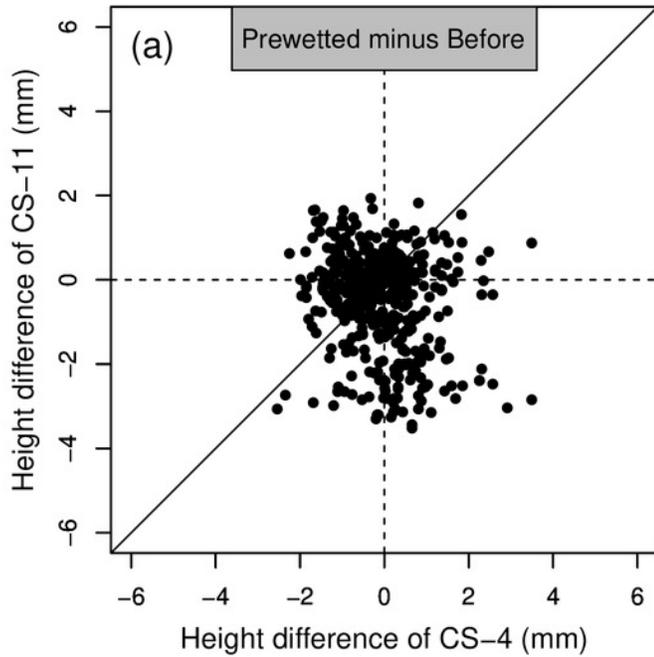
Figure 5

The height distribution on the four subplots of the conventionally farmed soil (CS) and organically farmed soil (OS) Before, Prewetted and After the prolonged rainfall events.

Four subplots over ten replicates combined, namely $n = 120$ for the overall, $n = 40$ for Before, Prewetted and After.
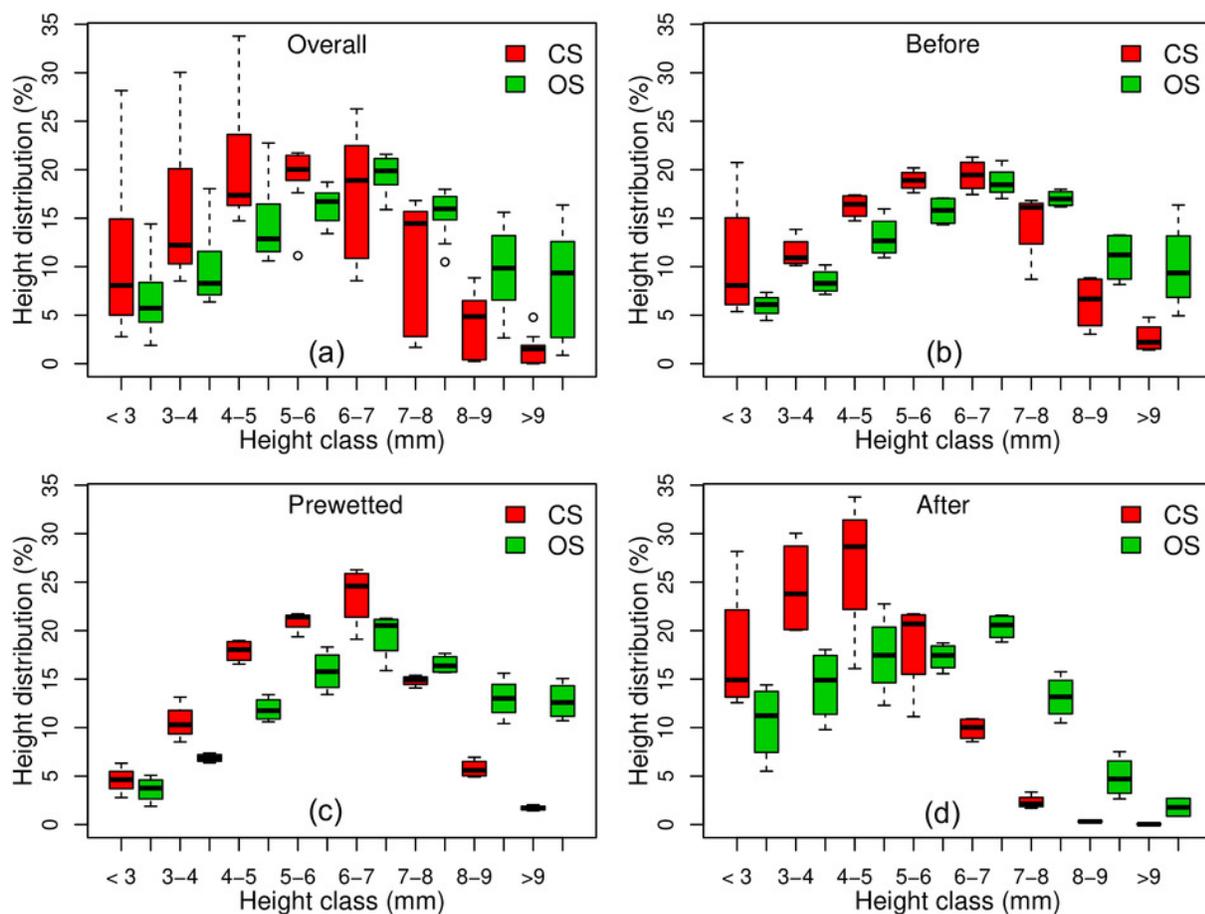
Figure 6

The surface roughness changes on the subplot b (as an example) of the conventionally farmed soil (CS) and organically farmed soil (OS) Before, Prewetted and After the prolonged rainfall events.

The data was combined from the subplot b over the ten replicates.
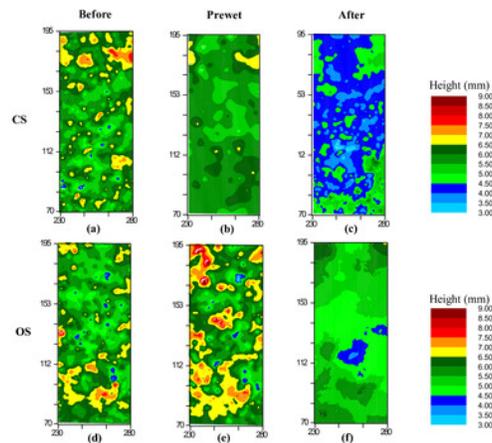
Table 1(on next page)

Selected properties of the conventionally farmed soil (CS) and organically farmed soil (OS).

Different superscripted letters in each column indicate significant differences (t-test). The subscripted numbers after each average value show the standard deviation ($n = 10$). For more soil properties, please refer to Hu et al. (2013).

	Clay (%)	Silt (%)	Sand (%)	Percentage of aggregates > 250 μm (%)	Soil organic carbon (mg g^{-1})
CS	16.8 ^{a} $_{1.38}$	71.47 ^{a} $_{1.76}$	11.50 ^{a} $_{1.00}$	66.85 ^{a} $_{0.47}$	10.9 ^{a} $_{0.05}$
OS	14.39 ^{b} $_{0.52}$	75.84 ^{b} $_{0.56}$	9.77 ^{b} $_{0.38}$	77.76 ^{b} $_{1.87}$	16.9 ^{b} $_{0.10}$

1

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Table 2(on next page)

Selected soil erosional responses on the conventionally farmed soil (CS) and organically farmed soil (OS) over the 360 min simulated rainfall events.

The subscripted numbers after each average value show the standard deviation ($n = 10$).

1

	CS	OS
Time to initiate runoff (min)	60	120
Time to reach runoff steady-state (min)	180	240
Time to reach peak ERsoc (min)	150	330
Average runoff rate at steady-state (mm h^{-1})	$12.9_{\pm 0.2}$	$10.7_{\pm 0.2}$
Average peak ERsoc	$1.92_{\pm 0.12}$	$1.44_{\pm 0.05}$
Total runoff on average (mm)	$55.6_{\pm 9.1}$	$34.1_{\pm 6.0}$
Total soil loss on average (g)	$27.4_{\pm 5.0}$	$16.1_{\pm 3.0}$
Total SOC loss on average (mg)	$369.1_{\pm 85.1}$	$326.0_{\pm 59.1}$
Loose materials remained on dried flumes (g m^{-2})	$10.96_{\pm 3.01}$	$43.78_{\pm 11.40}$

2

Table 3(on next page)

The erosional responses of the least and most eroded replicates on the conventionally farmed soil (CS) and organically farmed soil (OS) over the 360 min simulated rainfall events.

		Total rainfall (mm)	Total runoff (mm)	Total soil loss (g)	ERsoc peak	Total SOC loss (mg)
Least eroded replicate	CS-4	184.55	39.00	15.53	2.01	239.16
	OS-9	182.20	22.55	12.20	1.38	239.70
Most eroded replicate	CS-11	181.18	66.49	37.33	1.72	453.65
	OS-12	188.37	42.95	22.37	1.46	441.84

1