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Catchment drainage network scaling laws found experimentally 1 in overland flow morphologies 2

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11 Key Points:

12	•	Unchanneled surface under spatially non-uniform rainfall shows the same scaling
13		structures as catchment
14	•	The power law exponents remain constant during the surface evolution

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15 Abstract

The scaling relation between the drainage area and stream length (Hack's law), along with 16 exceedance probabilities of drainage area, discharge and upstream flow network length are 17 well known for channelized fluvial regions. We report here on a laboratory experiment on 18 an eroding unconsolidated sediment for which no channeling occurred. Laser scanning was 19 used to capture the morphological evolution of the sediment. High intensity, spatially non-uniform 20 rainfall ensured that the morphology changed substantially over the 16-h experiment. Based 21 on the surface scans and precipitation distribution, overland flow was estimated with the D8 22 algorithm, which outputs a flow network that was analyzed statistically. The abovementioned 23 scaling and exceedance probability relationships for this overland flow network are the same 24 as those found for large scale catchments and for laboratory experiments with observable 25 channels. In addition, the scaling laws were temporally invariant, even though the network 26

dynamically changed over the course of experiment.

28 1 Introduction

Even with markedly different environmental and geological conditions, catchment 29 drainage networks have similar geometrical characteristics that take the form of power laws 30 [Rodríguez-Iturbe and Rinaldo, 1997; Rinaldo et al., 2014], as measured for different areas 31 [Hack, 1957; Mandelbrot, 1977; Tarboton et al., 1989; Rigon et al., 1996]. Hack's law [Hack, 32 1957] states that the upstream length (l, the longest flow path into each point) and drainage 33 area (A) are related via a power law scaling $(l = A^h)$ where the exponent h (Hack exponent) 34 was measured in the range of [0.5-0.7] for different river networks [Hack, 1957; Gray, 1961; 35 Mueller, 1972; Mosley and Parker, 1973; Montgomery and Dietrich, 1992; Maritan et al., 36 1996; Rigon et al., 1996, 1998], with an average value of about 0.58 [Willemin, 2000]. Also, 37 for the fluvial parts of landscapes, power-law relations with exponent ranges of [0.42-0.45] 38 and [0.5-0.9] were observed for the exceedance probabilities of drainage area and length, 39 respectively [Rodríguez-Iturbe and Rinaldo, 1997; Rigon et al., 1996; Crave and Davy, 40 1997; Paik and Kumar, 2011]. Different explanations of these power laws are available [Banavar 41 et al., 1999; Dodds and Rothman, 2000; Birnir et al., 2001; Banavar et al., 2001; Birnir 42 et al., 2007; Birnir, 2008; Rinaldo et al., 2014], including self-organized dynamic systems 43 [Bak et al., 1988; Rinaldo et al., 1993; Marković and Gros, 2014], invasion percolation [Stark, 44 1991] and minimum energy dissipation [Rodríguez-Iturbe et al., 1992]. 45

Catchment drainage networks are essentially static structures in the landscape, i.e., 46 their temporal evolution cannot be readily measured. On the other hand, laboratory-based 47 experimental geomorphology has a longstanding tradition [e.g., Schumm and Khan, 1971; 48 Flint, 1973; Mosley and Parker, 1973; Parker, 1977] and permits detailed and rapid investigations 49 of changes in surface morphology due to rainfall or overland flow [e.g., Crave et al., 2000; 50 Brunton and Bryan, 2000; Römkens et al., 2002; Hasbargen and Paola, 2003; Gómez et al., 51 2003; Pelletier, 2003; Turowski et al., 2006; Babault et al., 2007; Yao et al., 2008; Tatard 52 et al., 2008; Paola et al., 2009; Bonnet, 2009; Berger et al., 2010; Graveleau et al., 2012; 53 Rohais et al., 2012; McGuire et al., 2013; Reinhardt and Ellis, 2015; Sweeney et al., 2015]. 54 For instance, dynamic changes of a rill network in uncohesive sediment under a constant 55 uplift rate were observed by Hasbargen and Paola [2000]. In contrast, rill networks in a 56 cohesive sediment evolved along the previously generated rills [Bennett and Liu, 2016] due 57 to surface resistance. Singh et al. [2015] generated rill networks in a 0.5-m $\times 0.5$ -m experiment 58 under spatially uniform but temporally variable rainfall and constant uplift rate. They found 59 that the drainage area distribution was described by a power law with an exponent of 0.5. 60 Similarly, *Bennett and Liu* [2016] examined rill formation at the flume scale $(7 \text{ m} \times 2.4 \text{ m})$ 61 and found an exponent of about 0.5 for Hack's law. 62 In summary, geometrical characteristics of catchment drainage networks have a high 63

degree of similarity. These same characteristics are evident in channeled surfaces in laboratory
 studies. Here, we extend these studies by considering the flow network on an unchanneled
 sediment. Specifically, we measured the surface evolution of an unconsolidated sediment

under non-uniform rainfall and overland flow such that no (observable) rills were formed.

- ⁶⁸ However, the surface roughness produces a drainage network representation of the overland
- ⁶⁹ flow, which is then subjected to geometrical analysis.

70 **2 Experiment**

A 2-m \times 1-m erosion flume with 5% slope (Figure S1) was filled to a depth of 15 cm 71 with unconsolidated sediments that had a mean diameter of 0.53 mm (Table S1 and Figure S2, 72 where S refers to the Supporting Information). Non-uniform rainfall with an average of 73 85 mm h⁻¹ and Christiansen uniformity coefficient [Christiansen, 1942] of 26% was applied 74 (Figure 1h). The non-uniform rainfall ensured that the flume drainage network varied both 75 spatially and temporally due to non-uniform erosion of the initially planar surface. The flume 76 had an impermeable base and was drained by a single, 4-cm wide outlet (Figure S1), located 77 at (x = 0, y = 0). The sediment became fully saturated during the first 15 min of precipitation, 78 which was accompanied by a rapid elevation drop at the outlet during the first 5 min. A 3D 79 laser scanner, with about 4-mm resolution, was used to extract Digital Elevation Models 80 (DEMs) at 0.25, 0.5, 1, 2, 4, 8 and 16 h. More details of the experimental setup are available 81 in the Supporting Information. With the same design and precipitation distribution, another experiment was carried out at 10% slope with an average rainfall of 60 mm h^{-1} that lasted for 83 20 h. The results of this experiment, which are similar to those presented here, are included 84 in the Supporting Information (Figures S8-S12). 85

3 Results and Discussion

The elevation change during the experiment is shown Figure 1. The sediment elevation was measured from the outlet (z = 0). For convenience, we refer to the ranges $z \le 60$ mm and $z \ge 60$ mm as the downstream and upstream, respectively. Overall, the morphology evolution can be divided into two steps: (i) until t = 4 h, most of the variation occurred at the upstream end while the downstream end did not show any considerable evolution, and (ii) after t = 4 h, the downstream morphology propagates into the upstream.

To characterize the morphology, a network was generated based on the measured surface scans (Figure 1a-g) and precipitation (Figure 1h). Pit points were removed following *Planchon and Darboux* [2002]. Similarly to large scale river networks, the discharge distributions (*Q*) and drainage area (*A*) are computed via the D8 algorithm [*O'Callaghan and Mark*, 1984]:

$$Q_i = \sum_{j=1}^8 w_{ji} Q_j + R_i \Delta x \Delta y \tag{1}$$

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$$A_i = \sum_{j=1}^8 w_{ji} A_j + \Delta x \, \Delta y \tag{2}$$

where the summation over *j* refers to the eight cells surrounding the *i*th cell. The slopes from each cell (*i*) into each of the eight neighbor cells (*j*) were calculated, with flow directed along the steepest descent. The value of w_{ji} is unity if the cell *j* flows into cell *i*, otherwise it is zero. $R_i \pmod{1^{-1}}$ is rainfall intensity at cell *i* (Figure 1 h) and $\Delta x \pmod{2^{1/2}}$ mm) are the grid sizes in *x* and *y* directions, respectively.

The distribution of drainage area and discharge at different times are plotted in Figures 2 113 and S4, respectively. At t = 0.25 h (Figure 2a), four separate branches depicted by A, B, C 114 and D drained into the flume's outlet (x = 0, y = 0). Then, at t = 0.5 h (Figure 2b), branch 115 C joined B and branch BC was generated while a minor change in the network was evident 116 in the upper part of the network. After 1 h (Figure 2c), junction A became attached to BC 117 and the pathway denoted ABC was formed. At t = 2 h (Figure 2d), the area drained by ABC 118 inclined to the right side. Furthermore, branch D drained a greater proportion of the precipitation 119 as it assumed part of the upstream area previously drained by ABC. Finally at t = 4 h, the 120



Figure 1. Measured morphology (z) evolution during the 16-h experiment (a-g). Initially, the flume slope was 5%, with y = 0 the lowest elevation and x being the transverse direction. The flume drained at a single point, located at (x, y) = (0, 0). Due to the spatially non-uniform precipitation (h), the morphology changes increase from the left side (low precipitation rate area) towards the right (high precipitation rate area).



Figure 2. Drainage area (*A*) distribution determined using the D8 algorithm and the measured morphologies shown in Figure 1a-g. Initially, the flow paths, e.g., at t = 0.25 and 0.5 h, reflect the initial surface condition and central drainage point at the flume exit. The labels A-D identify the main drainage pathways, which coalesced with ongoing erosion over the course of the experiment. The impact of the higher-intensity rainfall on the right side of the flume is manifested in the main flow path, which moves to the

right side during the experiment (more details given in the text).



Figure 3. Sediment surface at t = 0 (a) and t = 16 h (b). The uncohesive sediment had a wide range of particle sizes. Smaller particles were preferentially eroded during the experiment, leaving the larger particles as shown in (b). The dynamics of this surface evolution are reflected in the changing drainage networks computed using the D8 algorithm (Figure 2).

network ABCD was generated (Figure 2e). At later times (t = 8 h and 16 h), the high flow 121 part of ABCD became more dominant and moved to the right (Figure 2f and g). Variations 122 in the drainage area network (and discharge network in Figure S4) mostly occurred in the 123 first 8 h of the experiment, similarly to the surface morphology. Changes were less rapid in 124 the second 8 h, although the main structure of the network was reinforced and some local 125 changes to the low-order pathways took place. The evolution of the downstream (Figure 1e) 126 started at the same time as the network (ABCD) was generated at t = 4 h (Figure 2e). The 127 network's width function was computed for each scan to quantify its temporal evolution 128 (Figure S5). 129

Even though the flow covers the entire surface and is continuous (except perhaps for 141 raindrop impacts), the D8 algorithm leads to its description as a network, which was considerably 142 reorganized during the 16-h rainfall duration (Figure 2). We recall that these networks do not 143 represent observable surface rills, but rather the drainage network derived from the surface morphology as captured by the surface scans. As shown in Figure 3, due to shorter erosion 145 time scales, the fine sediment particles are rapidly removed while the larger particles move 146 slowly down the surface [Hairsine and Rose, 1992a,b; Polyakov and Nearing, 2003; Sander 147 et al., 2011; Wang et al., 2014; Kim and Ivanov, 2014; Cheraghi et al., 2016; Lisle et al., 148 2017] or are not moved at all, resulting in a surface partially covered by motionless pebbles. 149 Therefore, the network evolution is a result of size-dependent sediment particle transport and 150 raindrop-driven rearrangement on the surface. 151

We next examine the statistical characteristics of the network. We first consider Hack's 152 law [Hack, 1957], which is a well-known metric used in analyses of large scale river networks 153 [Maritan et al., 1996; Rigon et al., 1996; Dodds and Rothman, 2001a]. For our case, the A-l 154 distribution was divided into 20 bins on a logarithmic scale. For each bin, the ratio between 155 consecutive average moments of length were calculated. The results are plotted in Figure 4 156 for the first to four moments of l (n = 1, 2, 3, 4). They show a validation of a finite-size 157 scaling framework for the distributions of l, in the form of $p(l|A) = l^{-\xi}F(l/A^h)$ where 158 $F(x) \to 0$ for $x \to \infty$ and $F(x) \to 0$ for $x \to 0$, analogous to large scale river networks 159 [Rigon et al., 1996]. The power-law relationship is maintained for at least two orders of 160 magnitude, with the scaling exponent h in the range of [0.54-0.6]. Upper and lower cutoffs 161 affecting the scaling range were expected. Lower cutoffs are basically the limits of detectability. 162 Upper cutoffs are associated with the maximum cumulative area or flow rate [Rigon et al., 163



Figure 4. Ratios of consecutive moments of the upstream length distribution (*l*) at any point within subcatchments of area (*A*) identified by steepest descent directions. The slope of the log-log plot is Hack's exponent (*h*) at different times (t = 0.25-16 h). The *A*-*l* distribution was divided into 20 bins on a logarithmic scale, with the n^{th} moment of (*l*) for each bin denoted by $\langle l^n \rangle$. The curves of higher moments (n > 1) are shifted vertically for the purpose of visualization.





¹⁶⁴ 1996]. Another experiment at 10% slope with an average rainfall of 60 mm h⁻¹ (Figure S11) ¹⁶⁵ showed a range of [0.51-0.55] for the Hack exponent (*h*). For both experiments, the Hack ¹⁶⁶ exponents agree with those found for large scale river networks [*Hack*, 1957; *Gray*, 1961; ¹⁶⁷ *Mueller*, 1972; *Mosley and Parker*, 1973; *Mueller*, 1973; *Montgomery and Dietrich*, 1992; ¹⁶⁸ *Maritan et al.*, 1996; *Rigon et al.*, 1996, 1998], which are in the range [0.5-0.7], yet with a ¹⁶⁹ measured mean of about h = 0.58 [*Willemin*, 2000] and an analytical value of h = 0.57¹⁷⁰ [*Birnir*, 2008].

The distributions of (computed) drainage discharge, drainage area and upstream length 171 are plotted in Figure 5. In Figure 5a, the flume discharge can be separated into low ($q \le 1.1 \times$ 172 10^4 mm h^{-1} , medium $(1.1 \times 10^4 < q < 3 \times 10^6 \text{ mm h}^{-1})$ and high $(q \ge 3 \times 10^6 \text{ mm h}^{-1})$ 173 sections. The low discharge region mostly covers the left of the flume (Figure S4) where the 174 precipitation rate is lower. The values of P(Q > q) for these regions do not change during 175 the network evolution (from 0.25 h to 16 h). For the medium discharge regions, a power-law 176 relationship $(P(Q > q) = q^{-\varphi})$ describes the exceedance probability with an exponent of 177 $\varphi = 0.49$. The high discharge area shows the most temporal variability, which corresponds to 178 the changes of the main streams (A-D in Figure S4). Since the D8 algorithm selects a single 179 adjacent down-gradient cell to receive water from a given cell, potentially the predicted flow 180 becomes more localized than in reality. Also, flow disturbances due to raindrop impact and 181 resulting mixing are not accounted for. 182

Due to spatial and temporal variations of precipitation in natural settings, the distribution 183 of drainage area and upstream length are more commonly used metrics for describing river 184 networks at large (spatial) scales. Even though in this study no rills formed, the distributions 185 of drainage area and upstream length under this shallow, overland flow cross a number of 186 scales characterized by power laws ($P(A > a) = a^{-\beta}$ and $P(L > l) = l^{-\psi}$) with $\beta = 0.47$ 187 and $\psi = 0.75$, respectively (Figure 5b and c). Furthermore, at 10% slope with an average 188 rainfall of 60 mm h^{-1} , exponents of 0.49, 0.47 and 0.71 were found for power laws describing 189 discharge, drainage area and upstream length distributions, respectively (Figure S12). These results are similar to large scale river networks [Mandelbrot, 1977; Tarboton et al., 1989; 191 Rigon et al., 1996; Dodds and Rothman, 2001a,b,c; Rinaldo et al., 2014]. In addition, the 192 values of these exponents are close to analytical results, $\beta = 1 - h$ and $\psi = \beta/h$, derived by 193 Maritan et al. [1996]. 194

The consistency between the laboratory results in Figs. 4 and 5, and results for catchment 195 networks [e.g., Rodríguez-Iturbe and Rinaldo, 1997] points to an underlying governing 196 principle operating at different scales, such as the principle of minimum energy expenditure 197 [Rodríguez-Iturbe et al., 1992] that applies at equilibrium conditions for river networks. Similarly, recent work (Smith 2018) on equilibrium landscapes showed that overland flows 199 minimized a Lagrangian function of kinetic and potential energies. For both potential (viscosity 200 dominated) and inviscid flows and for fixed boundary conditions, energy dissipation continues 201 monotonically until the steady flow configuration is achieved, i.e., energy dissipation is 202 a minimum [Lord Rayleigh, 1893]. The energy minimization principle has been shown 203 exactly (by re-parametrization invariance arguments, and in the small gradient approximation) 204 to correspond to the steady-state solution of the general landscape evolution equation in fluvial regions [Banavar et al., 2001]. Deriving scaling properties and self-organization in 206 optimal networks is therefore tantamount to analyzing the underlying equations if steady-state 207 solutions are sought. Laboratory-scale rill networks were also shown to evolve towards the 208 minimum energy expenditure [e.g., Gómez et al., 2003; Berger et al., 2010]. However, for 209 unchanneled morphologies, further investigation is needed since our results suggest (approximately) 210 time-invariant scaling laws for a rapidly eroding surface. 211

The dynamics of eroding surfaces and related overland flow (including raindrop impact) can be modeled via different approaches, from mechanistic models that consider coupled overland flow and soil erosion [e.g., *Nearing et al.*, 1989; *Hairsine and Rose*, 1992a,b] to catchment scale landscape evolution models (LEMs) [e.g., *Willgoose*, 1989; *Howard et al.*, 1994; *Perron et al.*, 2008; *Smith*, 2018]. LEMs, which predict channel networks at both the catchment and laboratory scales, are relevant to our experimental results. We emphasize that our experiment involves continuous overland flow on an unchanneled surface in contrast to channelized flow in a catchment. Nonetheless, characterization of the overland flow on the
measured morphology via the D8 algorithm results in a network that is geometrically similar
to a catchment drainage network. The D8 algorithm provides a network representation of
the overland flow driven by gravity. This representation is an approximation, but allows for
a direct comparison of the unchanneled surface morphology in our experiments with the
channeled networks found in catchments and in laboratory experiments.

These experiments support a notable extension of what was previously thought about 225 the kind of recursive features shown by channeled landscapes at much larger scales. Unchanneled 226 landscapes were thought to obey diffusive evolution. For splash-dominated erosion studied 227 here, the scaling structures were replicas of those occurring at orders of magnitude larger 228 scales. It is totally remarkable that the aggregation patterns are independent of the specific 229 sediment transport type in erosional patterns. Moreover, the temporal stability of the scaling 230 structures we measure here suggests that indeed the planar features of steady states are reached 231 almost immediately by erosional surfaces, as was speculated but never shown for real river 232 networks. We suggest that the results could provide a test case for LEMs, which are applicable 233 at both the laboratory [Sweeney et al., 2015] and catchment scales [Perron et al., 2009] on 234 the condition that channels are formed. In the above-mentioned network analysis of Banavar 235 et al. [2001], diffusion was ignored, although it is present in LEMs. Since diffusion effects 236 will tend to smooth surfaces in LEM predictions, we speculate that our results will prompt 237 additional investigations of the role of diffusion in these models. That is, it remains to be determined if the scale invariance uncovered in this work can be captured by LEMs. 239

240 4 Conclusions

An evolving unchanneled surface under a spatially non-uniform rainfall was statistically 241 characterized in the same manner as large scale river networks by converting the continuous 242 overland flow into drainage area and discharge networks. The measurements show that although the surface morphology and the corresponding overland flow network changed markedly 244 during the experiment, the system preserved Hack's law and power laws in distributions of 245 drainage area, length and discharge. More importantly, the exponents, the values of which 246 are identical to large scale river networks, remained in a narrow range despite the considerable 247 change in the surface morphology and the corresponding network structure. This work provides, 248 for the first time, experimental support for the self-similar organization of landscapes even 249 where observable rills or channels are not formed on the surface.

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