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1 **New constraints on the long-term debris-flow incision**
2 **law based on high resolution topographic data**

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6 **Key Points:**

- 7 • With a novel method we extract gradient and concavity of all first-order channels
8 from 38 mountain catchments with meter-scale topography
9 • Slope gradient scales non-linearly with ¹⁰Be erosion rate and concavity remain sta-
10 ble near 0.1, across a range of rock-type and climate
11 • Our results support an updated debris-flow incision model linking flow mass and
12 sediment delivery, and newly constrain 3 key parameters

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Abstract

In spite of recent development, state-of-the-art debris-flow incision models (DFIM) remain under-constrained, limiting their integration in landscape evolution model (LEM), and thus our understanding of catchment erosion dynamics, including their altitude distribution, their response to climatic and tectonic changes and drainage network reorganization. Here, we use a novel method to extract channel heads on high-resolution topographies of 38 catchments spanning two orders of magnitude of erosion rates, and constrain the gradient and concavity of hundreds of first order-channels. We show that this is consistent with a novel, refined analytical DFIM prediction and constrain the exponents relating incision E_{df} , slope gradient S_{ch} and debris-flow thickness H : $E_{df} \propto S_{ch}^{4.2 \pm 0.9} H^{0.8 \pm 0.6}$. We also provide constraints on the range of possible value for debris-flow erodibility, as a function of debris-flow frequency, and suggest further analysis on channel junctions and concavity. These results pave the way to an implementation of a DFIM in 2D-LEMs.

Plain Language Summary

Our understanding of the long-term coupling between climate and tectonics activity and their effect on mountain building and sediment production and export is mediated by the use landscape evolution model, mathematical models describing the competition between rock uplift and erosion processes. In spite of recent development, erosion caused by debris flows remains too under-constrained to be included in LEM. Here, we use a novel method to extract channel heads on high-resolution topographies of 38 catchments spanning two orders of magnitude of erosion rates, and constrain the morphology of channels in their initial reaches. We show that this morphologies match the prediction of an improved debris-flow incision model and constrain the exponents modulating the dependency of erosion on slope gradient and debris-flow thickness. We also provide constraints on the range of possible value for debris-flow erodibility, as a function of debris-flow frequency, another key parameter of this model, which open the way to an implementation of this debris-flow incision model in LEMs.

1 Introduction

By transforming bedrock into sediments and routing them throughout entire continent, erosion and transport processes play a significant part in Earth's dynamic, including its climate. Quantifying these processes and modelling them in landscapes evolution models (LEMs) is therefore a necessary step for exploring diverse questions in geosciences such as the role of climate and tectonics in driving mountain uplift (Molnar & England, 1990), the controls on mountain ranges altitude (Whipple et al., 1999; Egholm et al., 2013), the global cycles of various elements including carbon (France-Lanord & Derry, 1997). While the first LEMs only considered basic fluvial incision and hillslope diffusion (Kooi & Beaumont, 1996), newer ones can include more refined details such as unsteady overland water flow (Davy et al., 2017), stochastic landsliding (Densmore et al., 1998; Campforts et al., 2020) or clast tracking (Carretier et al., 2016), allowing for a finer representation of landscape dynamics. However, debris flows remain missing in LEMs even if they may dominate incision and morphology in upstream mountain channels (Stock & Dietrich, 2003; Hergarten et al., 2016; Neely & DiBiase, 2023). This imply potential bias in representing head water catchment dynamics, which might control drainage network reorganisation and ridge migration (Mudd & Furbish, 2005; Willett et al., 2014), mountain elevation (Whipple et al., 1999) or sediment fluxes changes in response to tectonic or climatic changes (Benda & Dunne, 1997). Preliminary studies have proposed a debris-flow incision model (DFIM) at the event scale (Hsu et al., 2008; McCoy, 2012) and its consequences for channel longitudinal profile (Stock & Dietrich, 2006). More recently McGuire et al. (2023) presented a relatively parsimonious DFIM able to reproduce the transition from constant slope colluvial channel to concave fluvial chan-

nel observed in the field (Stock & Dietrich, 2003; DiBiase et al., 2010; Struble et al., 2023). This DFIM essentially relies on a non-linear scaling between erosion rate E , debris-flow erodibility and thickness, k_{df} and H , and the local slope gradient, $E \propto k_{df} S^\alpha H^\beta$, as initially proposed by McCoy (2012). But for now, k_{df} , α and β are largely under-constrained, limiting DFIM insights and implementation in 2D-LEMs.

Past developments of LEMs have especially been made possible by the use of topographic data for characterizing the shape of landscapes. Relationship between slope gradient and drainage area reflect the balance between the erosive power of erosion processes and tectonic uplift (Tucker & Bras, 1998; Tucker & Hancock, 2010). For instance, the characterization of the slope/area relationship of fluvial channels and the measurement of knick-point migration rate, together with physics-based developments, has allowed to constrain the fluvial stream-power incision model (Lague, 2014). Nevertheless, the low resolution of former topographic data limited the analysis of debris-flow dominated portions of the landscape. In particular, defining where debris-flow channels begin has long been a problem in geomorphology (Montgomery & Dietrich, 1988). New possibilities were opened by airborne LiDAR digital elevation models (DEMs) providing meter scale resolution, and revealed that channels experiencing debris-flow incision exhibit a distinct slope/area relationship from fluvial channels Stock and Dietrich (2003). In the last years a few more studies combined high-resolution DEMs and ^{10}Be -derived erosion rates with the goals to understand more precisely how debris-flows control channel shape and ultimately to connect it with candidate DFIM. In the San Gabriel (SG) and San Jacinto (SJ) mountains and the Oregon Coast range, the drainage area of transition from near constant gradient associated with debris-flow to concave fluvial channel was observed to increase with erosion rates (DiBiase et al., 2012; Penserini et al., 2017; Neely & DiBiase, 2023). Though some studies initially thought the characteristic gradient of these channels was uncorrelated to erosion rates (Penserini et al., 2017) recent studies found a weak but clear increase of channel gradient with erosion rates, both in the SG, and a compilation of US catchments (Struble et al., 2023; Neely & DiBiase, 2023). Further, Struble et al. (2023) showed that these correlations between morphology and erosion rates could be reproduced to first-order by the 1D-LEMs combining a DFIM and SPIM (McGuire et al., 2023). Various combinations of parameters were equivalent in fitting the data as long as they respected $\alpha > 3\beta$ and we do not know how the DFIM parameters may vary with lithology or climate. Additionally, it was suggested that the upstream channel slope gradient may decouple from erosion rates, based on a US data compilation (Neely & DiBiase, 2023), a trend potentially existing in the SG data too but not consistent with the DFIM (Struble et al., 2023).

However, a common issue of all morphometric analysis until now is that they (i) used diverse, unjustified definition of the upstream channel boundary, and (ii) fitted the empirical framework of Stock and Dietrich (2003) to the bulk catchment slope-area data to derive debris-flow metrics, even though it is not consistent with existing DFIM. To address these two challenges we propose to individually extract all first-order channels from a compilation of 38 catchments with known erosion rates, thanks to a novel method allowing to identify channel heads in the colluvial domain (Lurin et al., 2023) and to compare their average morphology to direct prediction from an updated DFIM. Thus, we obtain unprecedented constraints on several key parameters of the DFIM opening new opportunities for the inclusion of DFIM into 2D-LEMs.

2 Analytical model of the slope gradient in channels carved by debris-flow incision

Here, we focus on first-order channel immediately downstream channel heads, thus neglecting fluvial erosion and the threshold factor reflecting a reduction in incision when debris flows approach rest. Therefore we can write the debris-flow erosion rate, as a function of slope gradient S and debris-flow depth H :

$$E_{df} = k_{df} F_{df} t_p S^\alpha H^\beta \quad (1)$$

116 With t_p the time of passage, k_{df} an erodibility coefficient depending on bedrock and flow
 117 properties, F_{df} the debris flow frequency, and α and β scaling exponent (McGuire et al.,
 118 2023). Combining empirical laws relating debris-flow volume, M , peak discharge and ve-
 119 locity as found by Rickenmann (1999), mass conservation and a channel width scaling
 120 with drainage area (i.e., $w = k_w A^b$), Equation 1 can be reformulated as a function of
 121 A mainly (McGuire et al., 2023). Thus, assuming steady-state, it allows to express chan-
 122 nel slope gradient as a function of U and A .
 123 However, in contrast to McGuire et al. (2023) we assume that, in first-order channels each
 124 debris flow transport all sediment previously delivered by hillslopes, consistent with *en-*
 125 *masse* failure at high gradient and low drainage (Prancevic et al., 2014; DiBiase & Lamb,
 126 2019). This implies, from mass conservation, that we have $M = \frac{UA}{f_s F_{df}}$ with f_s the sed-
 127 iment fraction of the flow. With these assumptions, we derive an expression for debris-
 128 flow first-order channel slope gradient, S_{ch} :

$$129 \quad S_{ch} = K^{-\frac{\beta}{3\alpha-\beta}} k_{df}^{-\frac{3}{3\alpha-\beta}} \left(\frac{U}{F_{df}} \right)^{\frac{c_2(3-\beta)}{3\alpha-\beta}} A_{ch}^{-\frac{3(1-c_2)+(c_2-b)\beta}{3\alpha-\beta}} \quad (2)$$

130 Where $c_2 = 0.8 \pm 0.05$ is an empirical exponent scaling debris-flow discharge with vol-
 131 ume constrained by the literature Rickenmann (1999), $b = 0.2 \pm 0.1$ for upstream bedrock
 132 channels (DiBiase & Whipple, 2011; McGuire et al., 2023; Neely & DiBiase, 2023), and
 133 K is a combination of parameters involved in characterizing debris-flow erosion, veloc-
 134 ity, discharge, channel width analogous to the one for SPIM (Lague, 2014; DiBiase & Whip-
 135 ple, 2011) (See supplementary information for a full derivation).
 136 Equation 2 suggest that α and β have a determining impact on the way uplift, debris-
 137 flow-frequency and drainage area control channel slope. It defines a concavity for debris-
 138 flow channels:

$$139 \quad \theta_{df} = \frac{3(1 - c_2) + \beta(c_2 - b)}{3\alpha - \beta} \quad (3)$$

140 and it predicts a power-law scaling between S_{ch} and uplift, or erosion rate at steady-state
 141 with an exponent e_U :

$$142 \quad e_U = \frac{c_2(3 - \beta)}{3\alpha - \beta} \quad (4)$$

143 This model also suggests that detailed analysis of debris-flow channel slope and concav-
 144 ity in natural catchments may provide strong constraints on these two parameters.

145 **3 Compilation of steady-state catchments with known erosion rates** 146 **and high-resolution DEM**

147 To constrain the analytical model presented above we need accurate constraints
 148 both on the first-order channel topography and on the steady-state catchment-average
 149 erosion rate. Thus we limit our analysis to catchments with (i) available 1-m resolution
 150 LiDAR digital elevation models (DEMs), (ii) available ^{10}Be denudation rate measure-
 151 ments and (iii) evidence of steady-state topography without glacial overprinting in the
 152 headwaters were we find first order channels (Table 1). Indeed, inclusion of criteria (iii)
 153 removed many catchments with glacial topography in the French and Swiss Alps, or basins
 154 from the Feather river reported by (Hurst et al., 2013) which are undergoing transient
 155 readjustment. Further, we compute Chi-Z plots for each basins and rejected a number
 156 of basins (from DiBiase et al. (2010, 2012)) with deviations from linearity and large dis-
 157 order metrics which suggest some type of drainage reorganisation or transient dynam-
 158 ics (Perron & Royden, 2013; Hergarten et al., 2016). As a result of this selection, we fo-
 159 cus on 38 basins in four different mountain range, the San Bernardino (SB) and San Gabriel
 160 (SG) mountains, the Oregon Coast (OC) range and the Valensole Plateau (VP) in France
 161 (Table 1). These basins have different erosion rates ranging from about 0,03 mm/yr to
 162 3 mm/yr, and were derived from the same method (Mudd et al., 2016), as reported in
 163 the Octopus database (Codilean et al., 2022). Data from DiBiase et al. (2012) is not in

Table 1. Information about the studied basins. References: (Binnie et al., 2007; DiBiase et al., 2012; Marshall et al., 2017; Penserini et al., 2017; Godard et al., 2020) for B07, D10, D12, P17 and G20, respectively, with basin code from source. Erosion rates are from the Octopus database except for the references with a *. θ and S_{ch} values are median values from all 1st order channels extracted in the basin, while 25th and 75th percentiles are in supplement. Abbreviation for range and lithology: SG: San Gabriel; SB: San Bernardino; OC: Oregon Coast; VP: Valensole Plateau. C/ Crystalline basement; T: Turbiditic sedimentary rocks; Co: Clast-supported conglomerate; MAP: Mean Annual precipitation, as reported in Neely and DiBiase (2023) for US sites and Godard et al. (2020) for VP.

Code	Ref	E (mm/yr)	1σ (mm/yr)	Area (km ²)	Channels	θ	S_{ch}	Range / Litho / MAP (m/yr)
SGB5	D10	0.1464	0.041	9.9	563	0.14	0.69	SG / C / 0.6-1.3
SGB9	D10	0.448	0.093	17.3	424	0.07	0.51	SG / C / 0.6-1.3
SGB6	D10	0.267	0.149	9.7	408	0.13	0.70	SG / C / 0.6-1.3
SG123	D10	0.099	0.02	3.2	97	0.07	0.47	SG / C / 0.6-1.3
SG206	D10	0.278	0.103	5.4	175	0.05	0.73	SG / C / 0.6-1.3
SG207	D10	0.253	0.058	6.5	313	0.04	0.72	SG / C / 0.6-1.3
SG140	D10	0.182	0.039	7.7	0	0.00	0.00	SG / C / 0.6-1.3
SG0809	D12*	0.434	0.059	3.8	158	0.08	0.74	SG / C / 0.6-1.3
SG0804	D12*	0.493	0.111	2.5	98	0.13	0.74	SG / C / 0.6-1.3
SG0805	D12*	0.343	0.058	3.2	98	0.07	0.58	SG / C / 0.6-1.3
SGB10	D12*	0.279	0.023	3.0	185	0.10	0.63	SG / C / 0.6-1.3
SG126	D12*	0.591	0.071	2.3	122	0.00	0.82	SG / C / 0.6-1.3
SG125	D12*	0.465	0.061	2.0	89	0.08	0.77	SG / C / 0.6-1.3
SG818	D12*	0.711	0.13	2.0	132	0.18	0.71	SG / C / 0.6-1.3
SGB7	D12*	0.253	0.054	2.3	133	0.09	0.72	SG / C / 0.6-1.3
SG0704	D12*	0.668	0.186	2.1	106	0.04	0.67	SG / C / 0.6-1.3
SGB2/SG124	D12*	0.15	0.05	3.8	225	0.17	0.43	SG / C / 0.6-1.3
SG0748	D12*	0.541	0.147	2.0	117	0.11	0.72	SG / C / 0.6-1.3
SG127	D12*	0.736	0.099	2.3	143	0.06	0.82	SG / C / 0.6-1.3
SG151/SG152	D12*	0.256	0.18	2.0	49	0.04	0.54	SG / C / 0.6-1.3
SG0708	D12*	0.435	0.068	1.9	94	0.08	0.66	SG / C / 0.6-1.3
SG0749	D12*	0.611	0.195	2.6	109	0.07	0.77	SG / C / 0.6-1.3
CRN503	P17	0.087	0.021	18.2	438	0.06	0.69	OC / T / 2.1-2.3
HadsCRN	P17	0.111	0.03	1.8	79	0.06	0.63	OC / T / 2.1-2.3
Trib1	M17	0.078	0.02	0.7	20	0.15	0.51	OC / T / 2.1-2.3
Trib2	M17	0.055	0.014	0.3	10	0.01	0.68	OC / T / 2.1-2.3
CRN501	P17	0.072	0.041	2.0	85	0.10	0.56	OC / T / 2.1-2.3
CRN502	P17	0.067	0.017	0.7	34	0.06	0.51	OC / T / 2.1-2.3
CRN500	P17	0.059	0.013	2.0	83	0.04	0.50	OC / T / 2.1-2.3
Bin-1	B07	1.94	0.54	1.1	30	0.09	0.64	SB / C / 0.6-0.8
Bin-2	B07	2.8	0.764	0.6	28	0.19	0.76	SB / C / 0.6-0.8
Bin-3	B07	1.176	0.257	2.4	119	0.11	0.85	SB / C / 0.6-0.8
Bin-4	B07	1.032	0.223	0.9	76	0.15	0.85	SB / C / 0.6-0.8
Bin-5	B07	1.134	0.254	0.3	24	0.14	0.89	SB / C / 0.6-0.8
Va-19-K	G20*	0.05	0.02	3.5	68	0.10	0.34	VP / Co / 0.7
Va-18-G	G20*	0.05	0.02	2.4	58	0.01	0.24	VP / Co / 0.7
Va-15-M	G20*	0.045	0.018	1.5	27	0.11	0.16	VP / Co / 0.7
Va-13-ON	G20*	0.03	0.012	1.4	49	0.19	0.29	VP / Co / 0.7

164 this database but followed a similar method and rates reported by DiBiase et al. (2010)
165 with the same method are within 10% of their Octopus counterpart. For VP the erosion
166 rates is not the results of a sand sample at the basin outlet, but of in-situ hillslope
167 denudation rates converted to a basin estimates through an extraction of hilltop curvature
168 (Roering et al., 2007; Hurst et al., 2013) and their variability derive from variability in
169 hilltop curvature.

170 The sites represent diverse climatic and lithological conditions. SB and SM have a dry
171 Mediterranean climate with sparse vegetation and lie on crystalline basement (weath-
172 ered granite). VP has a similar Mediterranean climate with slightly more developed veg-
173 etation and lie on young, gravelly conglomerate. OC has about twice more precipitation,
174 with developed coniferous forest over most slopes and lie over turbidite sequence with
175 siltstone and sandstone. In addition, direct field evidence of debris-flow activity and in-
176 cision in their upstream channel area exist for several selected catchments in the Ore-
177 gon Coast range and the San Gabriel and San Bernardino mountains (Stock & Dietrich,
178 2003; Kean et al., 2011, 2013; DiBiase & Lamb, 2019).

4 Methods

We use the Matlab TopoToolbox to analyse the topographic data (Schwanghart & Scherler, 2014). In order to reduce meter-scale irregularities, we applied a Wiener filter on the DEMs. In contrast to previous studies, instead of extracting the bulk statistics of slope gradient at low drainage area (DiBiase et al., 2012; Struble et al., 2023; Neely & DiBiase, 2023) we extract independently the slope gradient of individual first order channels and then analyze their statistics.

4.1 Retrieving the channelized domain

Debris-flow channels are the most upstream channels in a catchment. In order to study the first order debris-flow channels, we thus have to retrieve channel heads. Several automatic methods exist for automatically extracting channel heads from high resolution topographic data. We use the CO²CHAIN method (Lurin et al., 2023), which was developed specifically for high erosion-rate DEMs and was shown to work well on catchments with diverse topography and erosion-rates, without needing recalibration. After retrieving the localization of channel heads, we use a D8 routing algorithm to extract the whole channel network. Depending on catchment size, we retrieved from a few dozens to a few hundreds first-order channels (Table 1), allowing to derive significant statistics on S_{ch} . In each basin, 75% of the first-order channels maximal drainage area is below 10,000 and 30,000 m^2 which we take as a sign that overall all the topographic analysis are mostly upstream of the transition toward channel with a "large concavity" ($> 0.3 - 0.5$) considered dominated by fluvial processes landscape, typically between 0.1 and 1 km^2 in the studied regions (Neely & DiBiase, 2023).

4.2 Analysing slope gradient within first-order channels

For each catchment, we retrieve the initial slope of the debris-flow domain, i.e. the slope at channel heads, as well as the concavity of first-order debris-flow channels.

Slope measurement is greatly affected by local, meter-scale irregularities which can reflect the presence of boulders, local heterogeneities in the bedrock or tree uprooting. We consider the resulting slope variations as not representative of the long-term effect of debris-flow erosion. Therefore, we need to calculate slope at a scale which smooths these variations without erasing actual long term slope variations. We found that averaging slope over a 7 m moving window along the channel flowpath greatly reduced pixel-to-pixel slope variation without smoothing too much the channel profile (Penserini et al., 2017).

In order to calculate the concavity θ of single channel segments, we consider the following relationship between slope S and drainage area A :

$$S = S_0 A^{-\theta}$$

Therefore, we perform a linear regression between the logarithm of slope and drainage area to retrieve the concavity. Both for concavity and channel gradient we compute the median, 25th and 75th percentile of the distribution of each catchment.

5 Results

5.1 Initial slope of the debris-flow channels

Given that gradient jointly depend on E and A , we used the median value of concavity for each catchment and corrected the slopes found at channel heads to compute the expected gradient at $A = 10^4 m^2$. The normalized median of initial gradient increases with the erosion rate E (Figure 1a). Ignoring the four VP basins which seem to deviate from the main trend, we performed a bootstrapped linear regression on the log-

224 arithm values of S_{ch} and E , producing 50 samples of 34 $E-S$ pairs, randomly drawn
 225 with replacement from our dataset, and yielding:

$$S_{ch} = (0.78 \pm 0.03)E^{0.15 \pm 0.03}$$

226 with $R^2 = 0.54$ ($N = 34$).

227 5.2 First-order channel concavity

228 We found median concavities slightly above zero for most catchments (Figure 1b).
 229 Despite smoothing gradient along the channels, concavities distribution remain broad
 230 with a number of convex channels (steepening downstream) and interquartile ranges typ-
 231 ically between 0.4 and 0.6 and apparently growing towards lower erosion rates most found
 232 in VP and OC catchments. Nevertheless, we do not observe a significant trend with ero-
 233 sion rate, and the average of median concavities from the 38 catchment seems well con-
 234 strained with $\bar{\theta}_{df} = 0.09 \pm 0.009$ (standard error of the mean with a standard devia-
 235 tion $\sigma = 0.05$).

236 5.3 Constraints on α , β and k_{df}

237 Our topographic analysis provides strong constraints on θ_{df} and e_U (Figure 1), which,
 238 together with literature constraints on b and c_2 constrain the possible value range for
 239 α and β (Figure 2 and Supplement for equations). Assuming uncorrelated, normal un-
 240 certainties we derive a best estimates with 1σ uncertainties of $\alpha = 4.2 \pm 0.9$ and $\beta =$
 241 0.8 ± 0.6 . The uncertainties are dominated by uncertainties on θ_{df} and we may derive
 242 a lower uncertainty range, $\alpha = 4.2 \pm 0.5$ and $\beta = 0.8 \pm 0.3$, if we assume that uncer-
 243 tainties on β and α only derive from the uncertainties on the mean concavity $\bar{\theta}_{df}$. This
 244 implicitly assumes that the variability of θ_{df} , within and between catchments (Figure 1),
 245 is not related to variations on α and β but to other aspects, such as fine-scale variabil-
 246 ity of c_2 and b , or measurement errors (from the DEM, the channel head identification
 247 or the extraction methods).

248 With these refined constraints on α and β and using typical values for the other
 249 physical and empirical parameters, as in McGuire et al. (2023) (see Supplementary for
 250 a full description), we may also derive constraints on two essential parameters of the debris-
 251 flow incision model, k_{df} and F_{df} (Eq 2, Figure 2b). Unsurprisingly we found a trade-off
 252 in which the best estimate of k_{df} (i.e., when $\alpha = 4.2$ and $\beta = 0.8$) decreases from about
 253 1 to 0.1 $m^{1-\beta} \cdot s^{-1}$ as F_{df} increases from 10^{-3} to $10^{-1} yr^{-1}$. However, uncertainties re-
 254 main very large (> 2 order of magnitudes) if we consider the range of possible values
 255 for α and β . Indeed, k_{df} tends to be proportional to $1/F_{df}$ when $\beta - > 0$ and also in-
 256 creases with decreasing α (See Supplementary). Again, if we limit uncertainty to the mean
 257 concavity θ and the propagated uncertainties shrink (Figure 2b).

258 6 Discussion

259 Before discussing our results in terms of channel morphology and implications for
 260 debris flow incision model (DFIM), we stress that the variability observed for different
 261 metrics may in part be due to the stochastic nature of debris-flow erosion. Indeed, even
 262 in a steady-state catchment, the (unknown) distribution of properties and timing of re-
 263 cent debris-flow may vary from channel to channel, leading to potential fluctuation of
 264 the morphometry around a mean prediction. Correlating morphometric variability to
 265 debris-flow history is a future research challenge beyond the scope of this study.

266 6.1 Slope and concavity of debris-flow channels

267 The relationship we found between initial slope and erosion rate is consistent with
 268 the latest studies conducted about debris-flow channels slope. While previous studies con-

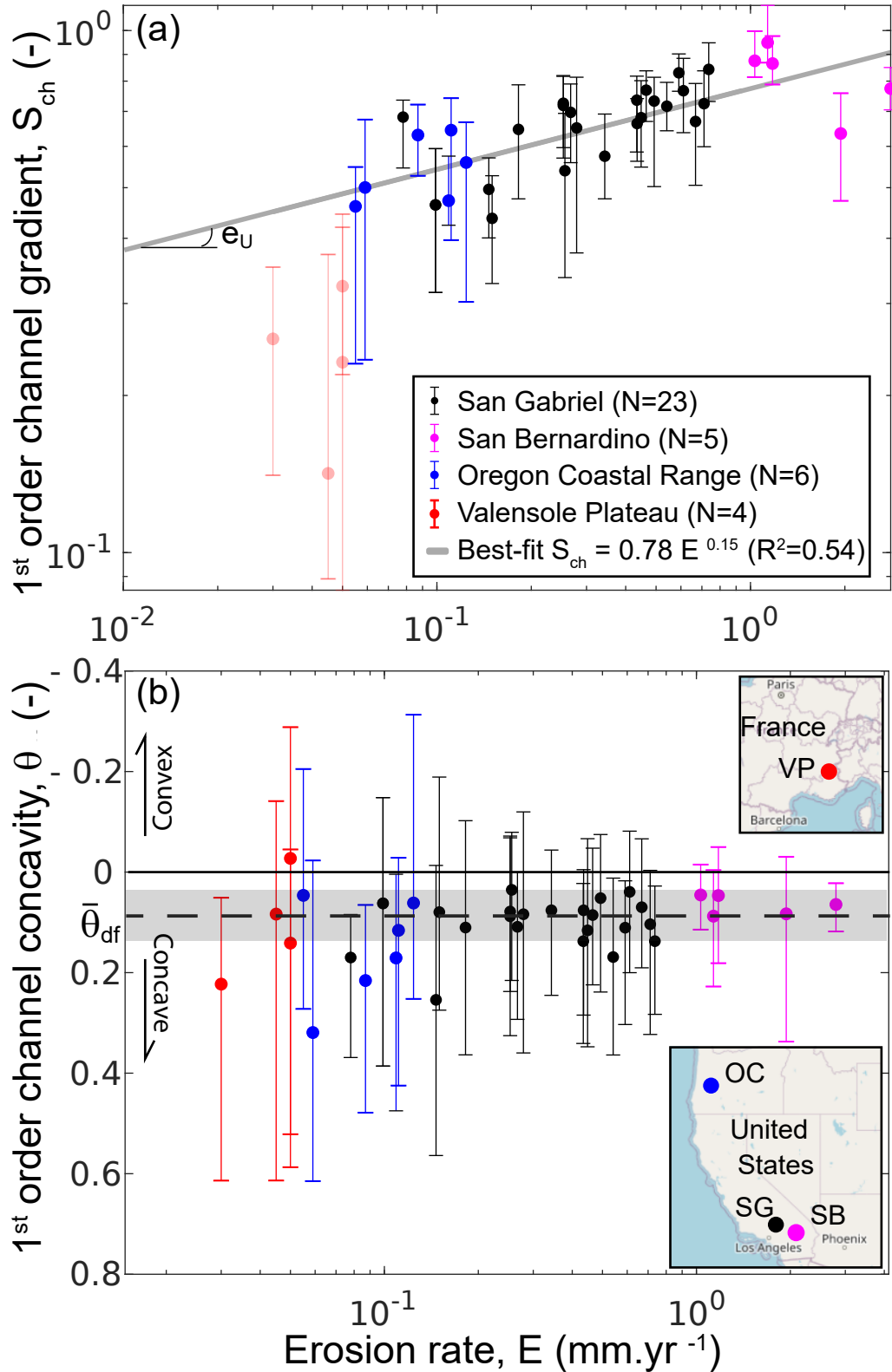


Figure 1. (a): Slope gradient of first-order channels (normalized at $A = 10^4 \text{ m}^2$) against erosion rate. (b): Concavity of first-order channels against of erosion rate. Dashed line and grey zone represent the mean concavity and 1σ range. Points with whiskers represent the median, and 25-75th percentiles of the distribution of all channels within basins color-coded by geographical areas. Insets show the location of the four geographical areas.

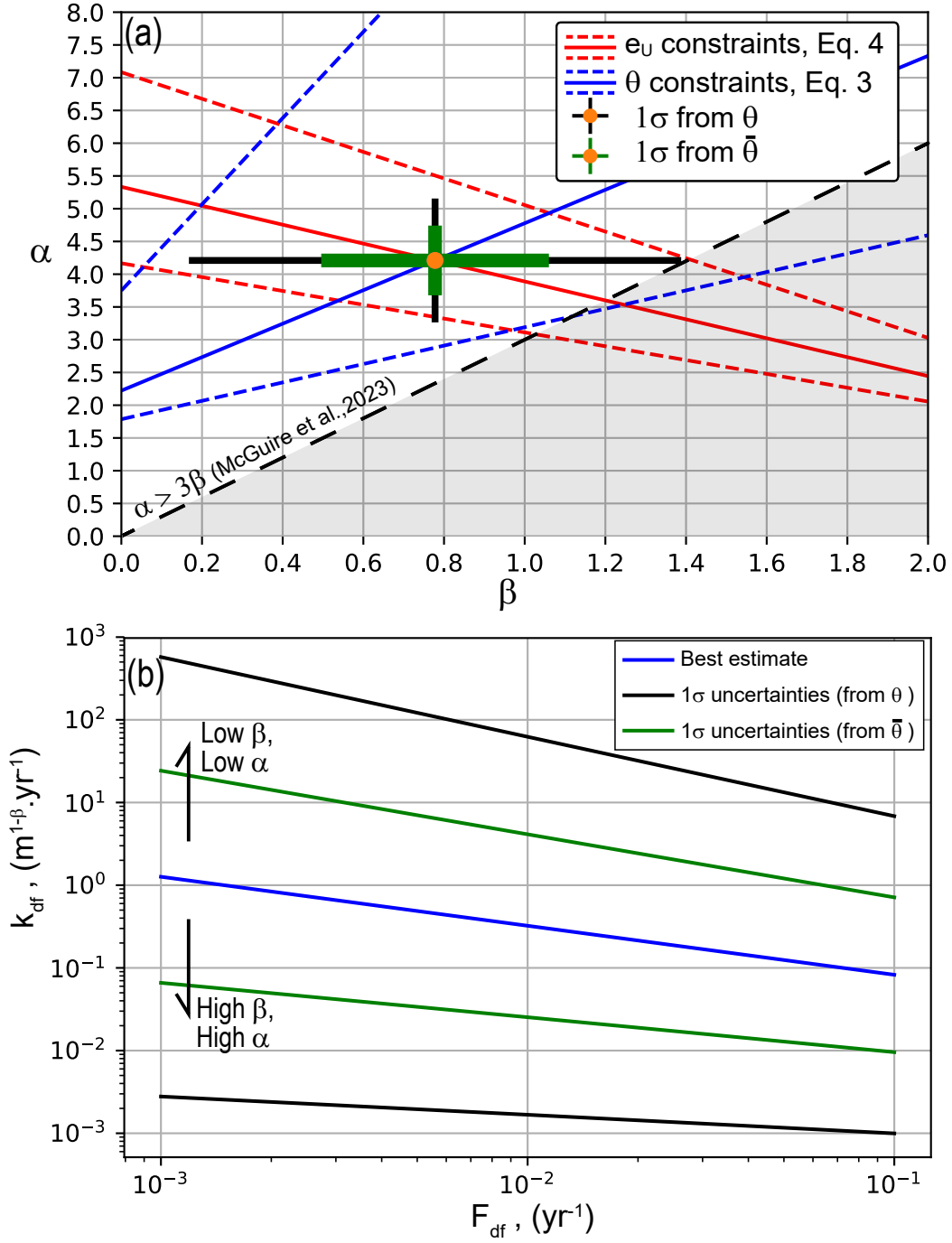


Figure 2. (a): Range of possible values for the α and β scaling exponents. Blue and red dashed lines are range of possible values when considering uncertainty range from the S-E scaling and concavity respectively. Black and green error bar are the 1σ uncertainties derived from Gaussian error propagation from variability on θ or $\bar{\theta}$. Dashed line and grey zone show the only pre-existing constraints before this study. (b): Range of possible values for K_{df} as a function of the debris-flow frequency, F_{df} , with error propagated from (a).

269 sidered the initial slope of debris-flow channels to be controlled by the threshold slope
 270 of hillslopes and therefore relatively independent on erosion rate (DiBiase et al., 2010;
 271 Penserini et al., 2017), recent work found a positive correlation between upstream chan-
 272 nel slope gradient and erosion rate (Struble et al., 2023; Neely & DiBiase, 2023). Struble
 273 et al. (2023) relationship between slope and erosion rate, obtained by fitting empirical
 274 debris flow scaling (Stock & Dietrich, 2003), to binned slope-area data from whole catch-
 275 ments in the San Gabriels Mountains is nearly the same as ours. However, our approach
 276 allowed a substantial scatter reduction and extended this result to a broader range of
 277 erosion rate and to other lithological and climatic settings. Indeed, our extended database
 278 demonstrate that the Oregon Coast Range and San Bernardin Mountains catchments
 279 follow the same trend, in spite of different lithology and climate, especially for Oregon.
 280 Based on data over several geographical zones, partly shared with our dataset, Neely and
 281 DiBiase (2023) proposed that debris-flow gradient could saturate near 0.8 above $E >$
 282 0.3 mm.yr^{-1} . Although less scattered (Figure S1) and more strictly selected, our dataset
 283 cannot completely disprove this alternative approach, given that the analytical DFIM
 284 predicts a weak power-law increase, and that small deviation for the two SB basins with
 285 the highest erosion rates may suggest saturation. Saturation may also reflect the inabil-
 286 ity of debris flows to compete with hillslope processes such as landsliding, as suggested
 287 by the observation that S_{ch} converge toward the mean hillslope gradient at high erosion
 288 rates (Figure S2). To distinguish these interpretations we suggest more analysis at high
 289 erosion rates ($\sim 1 \text{ mm/yr}$) and further attempts to understand the variability of S_{ch} .

290 As for the concavity of debris-flow channels, it has rarely been looked at (Lague
 291 & Davy, 2003), and recent studies on debris flow used the framework proposed by Stock
 292 and Dietrich (2003) which transition from fluvial concavity towards a null concavity at
 293 small drainage area. Interestingly, the concavity we found in the very first channel reaches
 294 is not null, and is quite consistent across two orders of magnitude of erosion rates and
 295 rather different environmental conditions (Figure 1). We recognize that measuring concav-
 296 ity at the scale of one single $\sim 100\text{m}$ long channel is difficult and that some of the
 297 intra-basin dispersion may be measurement errors (Figure 1(b)). Alternatively, part of
 298 the dispersion may also reflect local variability of controlling factors of the concavity, such
 299 as c_2 the volume-discharge scaling for debris flow, or b the channel width exponent. How-
 300 ever, the weak variations of the mean suggest that if these variations exist they remain
 301 around a fairly consistent mean, rather independent of uplift rate, climate and vegeta-
 302 tion. Another aspect meriting further research are the convex channels with $\theta < 0$ that,
 303 if correctly measured, would suggest unusual values for c_2 and b (e.g., $b > c_2$) or sug-
 304 gest that debris-volume do not grow linearly with A which may lead to initially convex
 305 channel (McGuire et al., 2023).

306 In contrast, the Valensole Plateau catchments appear to depart from the $E-S_{ch}$
 307 scaling, where they have a less steep gradient than other catchments. Nevertheless, concav-
 308 ity in VP appears consistent with the other catchments (Figure 1). Given that there
 309 is no documented debris-flow activity in these catchments, we cannot exclude that the
 310 low concavity of first order channels may result from other processes. For example in this
 311 catchment, sapping might be an important part of channel initiation and carving (Berhanu
 312 et al., 2012). Alternatively, the lower S_{ch} but similar concavity observed in Valensole catch-
 313 ments, may be due to a steady state driven by debris-flow incision with a higher value
 314 for debris-flow erodibility k_{df} or debris flow frequency, F_{df} than in the other basins. Given
 315 the weakly consolidated conglomerate, a higher debris-flow erodibility would be unsur-
 316 prising in these catchments.

317 6.2 Constraints on the debris-flow incision law

318 Our results are extending the analytical model developed by McGuire et al. (2023),
 319 modified here to consider that all sediments are moved by debris flows and to explicitly
 320 distinguish the role of debris-flow frequency and erodibility. We have shown that the mor-

321 phometry of first-order channel of catchments in various conditions appeared consistent
 322 with our updated DFIM (Eq 2) and yielded unprecedented constrains on the model ex-
 323 ponent. Preliminary modeling could suggest $\alpha > 1$ and $\beta < 1$ (McCoy, 2012), while
 324 further modeling and comparison with data clarified that $\alpha/\beta > 3 - 5$ (McGuire et
 325 al., 2023; Struble et al., 2023), but still considered very broad range of values typically
 326 $2 < \alpha < 8$ and $0.5 < \beta < 2$, and left untested values of $\beta < 0.5$. Although we do find
 327 a value of β near one and with a rather large range, including values below 0.5, we con-
 328 siderably narrow the possibility for α , and suggest $3 < \alpha < 5$.

329 Since debris-flow height is closely related to volume, the value of β controls whether
 330 several debris-flows are more or less incising than a single debris flow with equivalent to-
 331 tal volume. Our model predict that as long as $\beta < 3$, S_{ch} will increase with uplift
 332 rate (Fig 1) and thus will decrease with increasing debris-flow frequency, suggesting that
 333 frequent, small debris flows are more efficient than rare, large ones. Additionally, a po-
 334 tential way to further constrain β could be to analyse slope gradient change at the junc-
 335 tion of two first order channels to evaluate the effect of two different sources of debris-
 336 flow on one channel (Stock & Dietrich, 2006), considering that the two tributaries pro-
 337 duce debris-flow at the same frequency.

338 Narrower constraints on α and β could also be obtained by obtaining more constraints
 339 on θ_{df} , through additional topographic analysis and better understanding its variabil-
 340 ity in relations to b and c_2 . Interestingly, this requires studying debris-flow properties,
 341 and channel width and concavity evolution but does not require to constrain erosion rate
 342 and could thus be attempted in many more basins likely to be at steady-state. Such cou-
 343 pled analysis could also help reduce uncertainty on b and channel width evolution im-
 344 portant for both DFIM and SPIM (DiBiase & Whipple, 2011; Lague, 2014; Neely & DiB-
 345 iase, 2023).

346 Regarding k_{df} our observations yield a range of acceptable values anti-correlated to F_{df}
 347 which could be useful to guide future modeling studies. Physically, such scaling is con-
 348 sistent with the suggestion that in channels frequently scoured by debris flows the bedrock
 349 would have less time to weather in between events and thus be less erodible (Stock &
 350 Dietrich, 2006). Additional constraints on the model would benefit from a better under-
 351 standing of the control and magnitude of F_{df} . Importantly, F_{df} likely increases with the
 352 frequency of extreme climatic event that may induce debris flows, such as wild-fire or
 353 intense rainfall, (Godt & Coe, 2007; Kean et al., 2011, 2013; DiBiase & Lamb, 2019). How-
 354 ever, it is also likely that at landscape scale F_{df} increases with U , because of steeper hill-
 355 slopes and more frequent landsliding. Thus further improvement of DFIM could be gained
 356 by constraining long-term debris flow frequency, whether from sedimentological archive
 357 or other proxies (Stoffel, 2010; Ballesteros-Cánovas et al., 2024), and connecting it to cli-
 358 mate and tectonic conditions (Kiefer et al., 2021).

359 7 Conclusion

360 With a novel methodology for high resolution topography, we extracted all first or-
 361 der channels from a large compilation of steady state catchment with known erosion rates.
 362 We measured the initial gradient and concavity of these channels and found the former
 363 to increase as a power law of erosion rate and the latter to remain constant and posi-
 364 tive across two order of magnitude of erosion rates. These observations agree with the
 365 prediction of an updated version of the debris-flow incision model (DFIM) from McGuire
 366 et al. (2023) in which we enforce that all sediments delivered by the hillslopes are trans-
 367 ported by debris flows. Together, these observations determine optimal values for the
 368 exponent of the DFIM, $\alpha = 4.2 \pm 0.9$ and $\beta = 0.8 \pm 0.6$ with uncertainties dominated
 369 by the observed variability in concavity. Uncertainties may be halved if we only consider
 370 the variability of the mean concavity. Last, we provide first estimates of k_{df} as a func-
 371 tion of F_{df} that would match observed morphology.

372 Although additional constraints on other parameters of the DFIM (e.g., c_2 and b),
 373 the controls on debris-flow frequency F_{df} , and on other aspect of the morphology (e.g.,
 374 gradient reduction at channel junctions) are important challenges to address, the new
 375 insights on DFIM gained here pave the way for future studies with landscape evolution
 376 models (LEM) including DFIM. On the one hand the implications and prediction of ex-
 377 isting 1D-LEM (McGuire et al., 2023) could be substantially refined by exploring a re-
 378 duced parameter space. On the other hand, these constraints on the DFIM call for the
 379 development of new algorithm allowing to include DFIM in LEM, for example in mod-
 380 els already including stochastic processes such as landslides (Campforts et al., 2020), with
 381 the potential to offer new insights on steepland topography and its dynamics.

382 8 Open Research

383 All the DEMs used are available online. The San Bernardino Mountains, San Gabriel
 384 Mountains and Oregon Coast Range basins are found on the OpenTopography website
 385 (Milodowski, 2014; Heimsath et al., 2019; DOGAMI, 2011). The Haute Provence basin
 386 comes from the RGE-alti@database (<https://geoservices.ign.fr/rgealti>) (IGN, 2014). Ero-
 387 sion rates are available online from the reported literature or the Octopus database (<https://octopusdata.org/>).

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