Hydrological perspectives on precipitation intensity duration thresholds for
 landslide initiation: proposing hydro-meteorological thresholds

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12 ABSTRACT

13 Many shallow landslides and debris flows are precipitation initiated. Therefore, regional 14 landslide hazard assessment is often based on empirically derived precipitation-intensityduration (ID) thresholds and landslide inventories. Generally, two features of precipitation 15 16 events are plotted and labelled with (shallow) landslide occurrence or non-occurrence. Hereafter, a separation line or zone is drawn, mostly in logarithmic space. The practical 17 18 background of ID is that often only meteorological information is available when analyzing 19 (non-) occurrence of shallow landslides and, at the same time, it could be that precipitation 20 information is a good proxy for both meteorological trigger and hydrological cause. Although 21 applied in many case studies, this approach suffers from many false positives as well as 22 limited physical process understanding. Some first steps towards a more hydrologically based 23 approach have been proposed in the past, but these efforts received limited follow-up.

24 Therefore, the objective of our paper is to: a) critically analyse the concept of 25 precipitation ID thresholds for shallow landslides and debris flows from a hydro-26 meteorological point of view, and b) propose a trigger-cause conceptual framework for 27 lumped regional hydro-meteorological hazard assessment based on published examples and 28 associated discussion. We discuss the ID thresholds in relation to return periods of 29 precipitation, soil physics and slope and catchment water balance. With this paper, we aim to 30 contribute to the development of a stronger conceptual model for regional landslide hazard 31 assessment based on physical process understanding and empirical data.

33 1 INTRODUCTION

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35 Landsliding is one of the most abundant hazards having massive influence on socio-economic 36 functioning of society. Continuous development in mountain areas increases the exposure of 37 people and properties to the landslide hazards, with precipitation-initiated landslides being the 38 most common. On regional scale, the possibility of a landslide to occur can be assessed in 39 different ways (Chacon et al, 2006, for review): 1) heuristic, via susceptibility modelling; 2) 40 empirical, lumped-statistical, by relating rainfall information to the observed occurrence (e.g. 41 Caine, 1980; Wieczorek and Glade, 2005; Guzzetti et al, 2007; Guzzetti et al, 2008, and 42 reference therein); 3) by spatially distributed physical-deterministic modelling (e.g. Anderson 43 and Lloyd, 1991; Montgomery and Dietrich, 1994; Wu and Sidle, 1995; Borga et al, 1998; 44 Pack et al, 1998; Burton and Bathurst, 1998; Van Beek, 2002; Baum et al, 2008). The 45 heuristic models are mainly used in first assessments of (landslide) hazards for regional 46 planning. They are based on readily available static information, like topography, lithology 47 and land use, and then empirically related to historical landslide database (if available). The dynamic predisposing factors, like actual wetness state of the potentially unstable slopes, are 48 49 not taken into account. The physical process-based models can take into account the dynamics 50 of regional hazard assessment. Most of these models run spatially distributed hydrology -51 slope stability calculations, with different conceptualization and degrees of complexity for the 52 representation of the physical processes. Typically, the hydrology in these models at 53 catchment scale is not calibrated, or the calibration is restricted to the infiltration process or 54 local groundwater levels (if monitored). In such case, the correctness of the modelling is 55 assessed from how well local displacements or possible failure areas can be predicted. With 56 the increased availability of data and computational power, a range of these models has been 57 published with increased levels of complexity and applicability (e.g. Frattini et al, 2004; 58 Arnone et al, 2011; von Ruette et al, 2013; Lepore et al, 2013; Anagnostopoulos et al, 2015; 59 Aristizábal et al, 2016; Fan et al, 2016). However, the practical application of such 60 deterministic models, especially in terms of early warning systems, is still limited to specific 61 studies, due to the time effort and data demand.

The precipitation intensity-duration (ID) thresholds for landslide hazard assessment, however, see widespread application in early warning systems, both at local and regional scale. They are based on analysis of the dynamic variables precipitation and landslide occurrence, and require a high quality spatiotemporal landslide inventory and precipitation time series. Empirical-statistical precipitation thresholds are derived by plotting two 67 characteristics of precipitation, usually intensity (mm/hr or mm/day) and duration (hr or days), that have or have not resulted in landslides in a given area. The separation line, a 68 69 deterministic threshold or a probabilistic transition zone, between precipitation events 70 inducing landslides and events without hazards, is then drawn visually or by separation 71 techniques. Due to the spread of information over several orders of magnitude, it is usually 72 plotted in bi-logarithmic scale. Various precipitation ID thresholds for landslide initiation 73 have been derived for different physiographic settings and at various spatial scales (e.g. 74 Wieczorek and Glade, 2005; Guzzetti et al, 2007; Guzzetti et al, 2008; Peruccacci et al, 2017; 75 Rossi et al, 2017). The global and regional landslide precipitation ID thresholds encompass 76 different types of landslides and a distinct variety of geological and environmental factors, 77 such as lithology, soil depths and land use. The local ID thresholds are restricted more often 78 to relatively homogeneous conditions and mass movement types.

79 However, several shortcomings are frequently recognized and discussed. For example, 80 Berti et al (2012) recognized the problem of looking at landslide occurrence and disregarding 81 non-occurrence when applying the ID threshold. They used a Bayesian probability approach 82 to derive the probabilistic transition explicitly taking into account landslide occurrence and 83 non-occurrence. Also the role of hydrology in landslide initiation, although often 84 acknowledged to be of key importance, is usually not included in the statistical precipitation 85 ID threshold approach. Several attempts to more explicitly include hydrological predisposing 86 factors have been proposed mainly including measures for antecedent soil moisture content 87 (e.g. Crozier and Eyles, 1980; Glade et al, 2000; Godt et al, 2006; Ponziani et al, 2012) or by 88 splitting data sets in physiographic units like lithology, soil type, land use or season (e.g. Sidle 89 and Ochiai, 2006; Baum and Godt, 2010; Napolitano et al 2016; Peruccacci et al, 2017). 90 These approaches improved the predictive accuracy of the ID thresholds. However, to the 91 authors' knowledge, such studies have not been subject to a more thorough analysis of the 92 specific hydrological information needed for reliable local and regional hazard prediction.

93 Therefore, the objectives of this invited perspective are to: (a) critically analyse the 94 precipitation ID thresholds for shallow landslides and debris flows from a hydro-95 meteorological point of view; and (b) propose a conceptual framework for lumped hydro-96 *meteorological* hazard assessment based on the concepts of trigger and cause. We will frame 97 in this perspective some published examples and associated discussions, making reference to 98 work by colleagues who have already explored this avenue. The aim of this paper is to 99 contribute to the development of a stronger conceptual model for regional landslide hazard 100 assessment based on physical process understanding, not only on empirical data.

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2 HYDROMETEOROLOGICAL ANALYSIS OF ID THRESHOLDS

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103 COMPARING ID THRESHOLDS WITH IDF CURVES

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105 Both precipitation intensity-duration thresholds (ID) and precipitation intensity-duration-106 frequency curves (IDF) are empirical relationships linking the duration of a precipitation 107 event, D, with its average intensity, I=H/D, H being the precipitation depth during the event. 108 IDF curves are routinely used in stormwater and flood management design and predictions, as 109 they describe the relationship linking duration and mean intensity of precipitation events 110 characterized by the same return period (Chow et al., 1988). Several functional expressions 111 can be used to describe such a relationship (Bernard, 1932; Wenzel, 1982; Koutsoyiannis, 112 1998), most of which can be approximated, especially for durations longer than 1 hr, as a 113 power law:

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$$I = A \times D^B \tag{1}$$

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with B [-] being the slope of the log-plotted straight line and A [L/T] a measure of the rainintensity of a rain event of unit duration.

118 Equation (1) is also adopted to describe precipitation ID thresholds, the difference 119 being that the IDF curves are isolines of cumulative probability of precipitation events, 120 whereas the ID plots are empirical thresholds for shallow landslides and debris flow 121 occurrence. Figure 1 gives examples of IDF curves with a return period of 10 years from 122 different places around the world. A common feature of the curves is that, regardless of 123 geographic location, B ranges from -0.8 to -0.65 for rain durations longer than ~1 hour, while 124 it levels off to around -0.5 for $D \le 1$ hr for most IDF curves. Note that IDF curves are mostly 125 determined for rain durations up to 24 hrs. In the same graph, the upper envelope of the 126 largest precipitation values ever observed (World Meteorological Organization, 1986), is 127 plotted using the equation proposed by Brutsaert (2005), which has a smaller slope with B 128 equal to -0.52.



Figure 1. Examples of intensity-duration-frequency curves for 10 year return period (1-9) and curve of the
maximum observed precipitations (10). Location and source: 1 Najran region, Saudi Arabia (Elsebaie, 2012); 2
Uccle, Belgium (Van de Vyver, 2015); 3 Naples, Italy (Rossi and Villani, 1993); 4 Los Angeles, California
(Wenzel, 1982); 5 Pelotas, Brazil (Damé et al., 2016); 7 Hamada, Japan (Iida, 2004)); 8 Selangor, Malaysia
(Chang et al., 2015); 9 Sylhet, Bangladesh (Rasel and Hossain, 2015); 10 Greatest known observed point rainfall
(Brutsaert, 2005).





Figure 2. Rainfall intensity-duration (ID) thresholds. Numbers refer to case studies (Guzzetti et al, 2007). Very thick lines are global thresholds; thick lines are regional thresholds and thin lines are local thresholds. Black lines show global thresholds and thresholds determined for regions or areas pertaining to the Central to Eastern European region. Grey lines show thresholds determined for other regions or areas.

143 More than 90% of the landslides in the global data set are shallow landslides and 144 debris flows (Figure 2, Guzzetti et al, 2007). Note that the threshold is usually obtained as 145 lower envelope of the events resulting in landslide initiation, although also other threshold 146 definitions exist (e.g. Staley et al., 2013; Ciavolella et al, 2016; Peres and Cancelliere, 2016). 147 Obviously, ID thresholds differ greatly between climate and physiographic regions, especially 148 in absolute values. Therefore, scaled representations have been proposed for the thresholds, 149 such as dividing precipitation intensity by the mean annual precipitation in order to better 150 compare the thresholds (Guzzetti et al, 2007). However, in our analysis the focus is on the 151 unscaled measured precipitation ID representation, as it is a convenient way to compare with 152 IDF. The exponent of most of the reported thresholds for initiation of landslides range 153 between -0.2 and -0.6. For landslides triggered by short precipitation events (D \leq 1 hr), the 154 slopes of the IDF and ID curves substantially coincide (Figure 3).. On the other hand, for 155 longer precipitation durations, ID thresholds have smaller slopes than IDF curves. This means 156 that landslide initiation on the right side of the graph (lower precipitation intensity with longer 157 duration) would occur with rapidly increasing return periods of precipitation events. This is 158 counter-intuitive, as during a long-lasting wet period landslides are usually more frequent, 159 while many debris-flows triggered by very short and intense storm originate from channel bed 160 mobilization rather than being (new) mass movements. This shows that the method used to 161 derive ID thresholds for landslide initiation based on landslide and precipitation reports leads 162 to troublesome interpretations. Owing to the high spatial variability of rainfall at scales 163 smaller than 5 km (e.g. Krajewski et al., 2003; Ciach and Krajewski, 2006) and the limited density of operational rain gauge networks, the rainfall intensity observed by rain gauges 164 165 systematically underestimates the actual triggering rainfall intensity at debris flow initiation 166 locations (Marra et al., 2016), especially for short rain duration and high return period (Destro 167 et al., 2017). This issue has been shown to significantly affect the obtained ID thresholds for 168 debris flow initiation (Nikolopoulos et al., 2015; Marra et al., 2017). Additionally, different 169 methods adopted to define the dry period separating rain events have been shown to strongly 170 affect the ID threshold (e.g. Vessia et al., 2014; Melillo et al; 2015). Furthermore, several 171 authors already pointed out that characterizing a storm with its mean intensity, thus neglecting 172 peaks and underestimating actual intensity, affects the estimated probability of landslide 173 occurrence (e.g. D'Odorico et al., 2005; Peres and Cancelliere, 2016), and such an issue is 174 obviously more significant for long storm durations. In fact, for the rainfall depth data used to 175 derive IDF curves, the considered duration is simply a moving interval along the rainfall time 176 series, regardless from the actual beginning and end of a rainfall event. So, the corresponding

177 mean intensity usually refers, especially for short durations, to the heaviest part of a longer 178 rainfall event. Conversely, whatever the criterion adopted for the definition of a rainfall event, 179 in the case of ID threshold curves, the plotted mean intensity refers to the entire rainfall event. 180 Thus, within very long events leading to landslide triggering, there is very likely an intensity 181 peak, which is the "real" landslide trigger, preceded by a period of rain which contributes to 182 predispose the slope to failure. In the (D, I) plane, the point corresponding to the peak would 183 be shifted left- and upwards, compared to the point of the entire event. Given the typical 184 slopes of IDF and ID threshold curves, this shift likely corresponds to a smaller return period. 185



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Figure 3. Schematic representation of precipitation IDF curves, isolines of accumulated precipitation (ΣP) and ID
 threshold for shallow landslides and debris flows (simplified from Figure 2).

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191 HYDROLOGICAL INTERPRETATION OF ID THRESHOLDS

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193 The precipitation ID thresholds are *"volumetric*', i.e. every point depicts a total, cumulative 194 amount of precipitation. In Figure 3 the global summary of ID thresholds for shallow 195 landslides and debris flows (Guzzetti et al, 2007) is schematically represented by the dark 196 grey area, but, added to it, are isolines of accumulated precipitation volume (1, 10, 100 and 197 1000 mm). The first observation is that the regional and global landslide thresholds clearly 198 follow a slope different from isolines, meaning that longer duration landslide triggering 199 thresholds require larger water volume. This is understandable if landslides will be deeper 200 seated. However, the database contains mainly shallow landslides and debris flows (Guzzetti 201 et al, 2007). Clearly, this indicates the role of hydrology, or, to be precise, the balance 202 between infiltration, storage and drainage capacity of a slope (Bogaard and Greco, 2015).

203 Many of the reported empirical precipitation thresholds range between 10 and 100 mm 204 of accumulated precipitation. However, also <10 mm and >1000 mm volumes needed for 205 landslide initiation have been reported (e.g. as summarized in Guzzetti et al 2007). Under 206 'normal' antecedent wetness conditions (that is, soil field capacity), an accumulated 207 precipitation of < 10 mm is generally not capable of triggering a landslide or debris flow. Of 208 course, such an accumulated precipitation volume can trigger a shallow landslide or debris 209 flow in case of nearly saturated antecedent conditions. In this latter case, the reported 210 precipitation event is really the last 'push', the so-called trigger (see next section). On the 211 other hand, precipitation volumes > 1000 mm and/or durations of over 100 or even 1000 212 hours (> 1 month) are difficult to interpret in terms of average precipitation intensities and 213 triggering thresholds for shallow landslides and debris flows. Our point here is that the current 214 ID concept incorporates an unacceptable wide range of information with different types of 215 hazards (debris flows and landslides related to different hydrological processes), different 216 temporal meteorological information (from minutes to several days). This makes the use of ID 217 thresholds cumbersome or even misleading.

218 Additionally, ID thresholds have been derived by applying physically-based models of 219 infiltration and slope stability evaluation, which account for soil hydraulic properties, 220 different initial moisture conditions and the boundary conditions through which the slope 221 exchanges water with the surrounding hydrological system (e.g. Terlien, 1998; Rosso et al., 222 2006; Salciarini et al., 2006; Frattini et al., 2009; Papa et al., 2013; Peres and Cancelliere, 223 2014). Such physically-based thresholds often do not follow equation 1, generally adopted for 224 ID thresholds. For long precipitation durations, the physically-based ID curves tend to flatten 225 (e.g. Rosso et al., 2006; Salciarini et al., 2006), indicating that landslide initiation thresholds become less sensitive to (average) precipitation intensity, which is counter-intuitive and a 226 227 poor explanatory variable for landslide initiation.

Interestingly, Frattini et al. (2009) followed an inverse approach and obtained estimates of the probability of the precipitation characteristics leading to shallow landslide initiation by also considering antecedent precipitation. In particular, they showed how the 231 exponent of the IDF curves of their study area (a catchment located on the east side of the 232 Lake Como, in Lombardy, northern Italy) changed from -0.65, for unconditional probability 233 of triggering events, to -0.43, when 300 mm of rainfall in the previous four days was 234 included, thus approaching the slope of the observed ID thresholds. Antecedent precipitation 235 can be seen as an indirect means to account for the moisture conditions of the soil cover 236 before a triggering event. Therefore, the results of Frattini et al. (2009) can also be interpreted 237 as an indirect confirmation that considering the integral hydrological processes would 238 improve the performance of landslide initiation thresholds.

239 Greco and Bogaard (2016) give an example of the possible inclusion of slope 240 hydrological processes in the definition of landslide initiation thresholds for the case of a 241 slope covered by loose granular volcanoclastic deposits overlying a fractured limestone 242 bedrock. The hydraulic characteristic curves of the volcanic ashes constituting the majority of 243 the soil cover were known (Damiano et al., 2012; Greco et al., 2013), as well as the moisture 244 state of the cover before all 78 observed rainfall events (Comegna et al., 2016). Hence, it was 245 possible to define non-dimensional variables comparing the meteorological triggers with the 246 infiltration and storage capacity of the soil cover. This non-dimensional hydro-meteorological 247 threshold performed slightly better than the precipitation ID threshold in separating events 248 resulting in factors of safety smaller and greater than 1.3. The choice of referring to a factor of 249 safety larger than 1.0 was dictated by the actually observed soil conditions during the 250 monitoring period.

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252 3 TRIGGER - CAUSE CONCEPT: PROPOSING HYDROMETEOROLOGICAL253 LANDSLIDE THRESHOLDS

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255 In the strict sense, the precipitation ID threshold is an empirical-statistical threshold drawn to 256 separate failure and non-failure conditions based on observed landslides and precipitation 257 records. Precipitation is described in terms of average intensity and duration. The main 258 assumption is that there is an underlying causal relation between the recorded precipitation 259 event and the landslide occurrence. However, by including durations up to e.g. 1 month, the 260 direct causal relationship is weak and the method implicitly includes the wetness state of a 261 region. This limitation has been recognized from the start of using ID thresholds. For several regional hazard assessment analyses, research groups have extended the ID threshold method 262 263 by replacing the duration of a precipitation event on the x-axis with an antecedent 264 precipitation index or accumulated rainfall over a certain time interval (e.g. Crozier and Eyles,

265 1980; Glade et al, 2000; Chleoborad et al, 2006; Chleoborad et al, 2008; Scheevel et al, 2017). 266 This, however, leads to limited added information as still only precipitation information is 267 used. On the contrary, by replacing the x-axis with a direct measure or proxy for antecedent 268 soil water content, physically relevant information is added (e.g. Crozier and Eyles, 1980; 269 Wilson 1989; Wilson and Wieczorek, 1995; Crozier, 1999; Glade, 2000; Chirico et al. 2000; 270 Gabet et al, 2004; Godt et al, 2006; Ponziani et al, 2012). Interestingly, by including a water 271 balance of the potentially unstable soil, a statistical ID threshold evolves conceptually from a 272 plot with one prevalent driver and data source (precipitation) into a plot containing two 273 predominant drivers with two distinct time-scales: the antecedent hydrological 'cause' and the 274 precipitation 'trigger'. Besides soil water balance calculations, different sources of 275 hydrological information can be used to quantify the hydrological 'cause' of landslides. This 276 is largely unexplored ground, partly as data availability can be cumbersome and partly 277 because physically-based, (semi-) distributed modelling was preferred.

278 Concerning the 'trigger'-axis, there is little debate; it is the rainfall intensity 279 responsible for the short-term last push initiating a landslide. The time-scale for local and 280 regional assessment of course depends on the local situation, but hourly or daily time-scales 281 are the most common. The 'cause'-axis should represent the predisposing condition of the 282 area under study. For hydrologically triggered landslides, it should be related to the 283 antecedent wetness state of the area. However, there are several possible choices of 284 hydrological variables to be plotted along the 'cause'-axis, such as (effective) soil water 285 content, relative catchment storage and representative regional groundwater level. The choice 286 for the 'trigger' and 'cause' also implies a definition of the time-scale separating trigger from 287 cause, which should be related to the characteristics of the triggered landslide, but is in 288 practice often limited by the (temporal resolution of the) available data.

289 As mentioned before, there are - besides the soil moisture storage calculations 290 previously described - various examples of hydrological information added to landslide 291 thresholds. Hashino and Murota (1971) published an analysis of landslide triggers in a 292 catchment related to debris production using measured river discharge data to link the 293 landslide triggers to the water balance of the catchment. They identified that the landslides in 294 their study area occurred during above average antecedent conditions. This is one of the 295 earliest reported studies we know of explicitly looking at catchment water balance as an 296 important source of information on the antecedent hydrological condition of an area in 297 relation to landslide occurrence. Reichenbach et al (1998) made a combined flood and 298 landslide hazard analysis of the Tiber river catchment using 72 years of historical daily

299 discharge data from different gauging stations where hydrological parameters were calculated, 300 such as maximum mean daily discharge, specific discharge and flood volume and duration. 301 Probability of occurrence of landslides and floods was based on the ranking of the events. 302 Combining maximum mean daily discharge and discharge intensity, regional hydrological 303 thresholds for landslide and flood hazard (individually or combined) could be drawn. Chitu et 304 al (2016) followed a somewhat similar approach, analyzing the river discharge in several 305 catchments in the Ialomita Subcarpathians in Romania for landslide events in 2014. The 306 catchments could be characterized as having low/high relative storage. Additionally, a 307 calibrated regional rainfall-runoff model was used for hydrological analysis of landslides in 308 specific catchments. Detailed analysis of the (modelled) hydrological response indicated that 309 in two catchments with low infiltration capacity the direct runoff was strongly related to 310 landslide occurrence, whereas it could be linked to modeled soil infiltration flux in another 311 catchment. Extending the above to deep-seated landslides, the connected regional 312 groundwater level could be informative. Bogaard et al (2013) studied the hydro-313 meteorological triggering threshold of the re-activating coastal Villerville-Cricqueboeuf 314 landslide, Normandy, France. In this situation the hinterland of the coastal cliff consists of a 315 well-defined regional groundwater level. Landslide reactivation was seen to take place only 316 when water level was in the upper, more permeable top layer. The triggering rain event 317 together with surpassing a certain regional groundwater threshold could explain 3 of the 4 re-318 activations. Note that these groundwater levels were not taken in the active landslide area but 319 several kilometers inland.

320 Recently, further attempts have been made to use river discharge and lumped water 321 storage in a catchment as a proxy for the predisposing conditions for landslides along its 322 slopes. Following Hashino and Murota (1971), the basic idea is that when 'more-than-323 average' water is stored in the catchment, it is more likely that a rainfall event triggers 324 landslides. The disadvantage of using catchment wide storage is the relatively low spatial 325 resolution, and the difficulty of having (reliable and homogeneous) discharge time series in 326 catchments. Moreover, catchment storage assessment needs information on evaporation which 327 can have significant uncertainties. Of course, such an approach works only if the causes of the 328 predisposing conditions for landslides are somewhat related to catchment scale hydrological 329 processes. Ciavolella et al. (2016) defined a cause-trigger hydro-meteorological threshold in 330 the catchment of river Scoltenna, in Emilia Romagna (Italy), linking catchment storage and 331 event rainfall intensity, and compared its performance with that of a statistical ID 332 precipitation threshold. The two thresholds performed similarly, with the hydro333 meteorological thresholds being more accurate for identifying landslides, but giving a334 somewhat larger number of false positives.

These examples indicate that considering hydrological causes could be useful for a better identification of landslide initiation, but, at the same time, they show that the correct identification of the hydrological processes involved in the establishment of the predisposing conditions for landslides is mandatory for choosing the most informative hydrological variable to be plotted along the x-'cause'-axis.

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4 CONCLUDING REMARKS AND OUTLOOK

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343 The intrinsic limitations of precipitation ID thresholds for the identification of landslide 344 initiation conditions has been noted for some time. Indeed, such thresholds neglect the role of 345 the hydrological processes occurring along slopes, which predispose hillslopes to failure 346 (causes) and focus predominately on the characteristics of the last rainfall events leading to 347 slope failure (triggers). As a consequence, the predictive accuracy of the ID thresholds is 348 often low, even when they refer to small areas. We argue that the threshold values for rainfall 349 intensity of short and long duration (the far left and right side of the graphs) have limited 350 physical meaning and, consequently, that the use of precipitation ID thresholds can lead to 351 misleading interpretations of initiation conditions, as important antecedent conditions and 352 rainfall intensity variations are not taken into account. For this reason, we here advocate to be 353 very careful in uncritically using the precipitation ID thresholds as kind of regional 354 characteristic of (shallow) landslide occurrence.

355 Equally, for this and several other reasons, many colleagues advocate the use of 356 spatially-distributed physically based models for assessing landslide probability. The obvious 357 downside is that large data input and a well calibrated model are required. However, it is fair 358 to say, data are becoming more and more available and even precipitation predictions are 359 improving rapidly, especially with short lead-time. The use of high quality rainfall prediction 360 with very short lead time (e.g. 3 hours) requires efficient numerical models combined with 361 high computational power, especially if predictions are used for early warning purposes. This, 362 in practice, is still easier said than done. Therefore, we believe, that lumped, empirical (or 363 semi-empirical) thresholds will continue having a practical value, which still justifies 364 scientific attention.

We propose to use the cause-trigger concept for defining regional landslide initiation thresholds. This, we agree, is challenging, but, in our opinion, not impossible. First of all, it is 367 needed to define the characteristic time-scale separating the (dynamic) long-term predisposing 368 hydrological cause from the short 'final' landslide hazard triggering. This obviously depends 369 on the landslide type and physiographic characteristics. Looking at the discussed examples, it 370 becomes clear that the choice of the most informative hydrological variable to be used as a 371 proxy for landslide predisposing conditions strictly depends on site-specific geomorphological 372 characteristics, and that accurate analysis of the boundaries through which the potentially 373 unstable area exchanges water with the surrounding hydrological systems is mandatory. In 374 other words, for the assessment of landslide predisposing conditions, the water balance of the 375 slope should be assessed, but getting the information about the inherent hydrological 376 processes (e.g. evaporation, runoff, groundwater recharge) at the required spatial-temporal 377 resolution is often a challenge and could require some kind of calculations or modelling. 378 However, rapidly more and higher resolution hydrological data become available which can 379 be used in assessing the hydrological predisposing condition.

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