



1 Invited perspectives. A hydrological look to precipitation intensity duration
 2 thresholds for landslide initiation: proposing hydro-meteorological thresholds

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 12 **ABSTRACT**

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

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

33



34 1 INTRODUCTION

35

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

63 The precipitation intensity-duration (ID) thresholds for hazard assessment, however,
 64 see widespread application in early warning systems, both at local and regional scale. They
 65 are based on analysis of the dynamic variables precipitation and landslide occurrence, and
 66 require a high quality spatiotemporal landslide inventory and precipitation time series.
 67 Empirical-statistical precipitation thresholds are derived by plotting two 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 deterministic threshold or a probabilistic transition zone, between precipitation events inducing landslides and events without hazards, is then drawn visually or by separator techniques. Due to the spread of information over several orders of magnitude, it is usually plotted in bi-logarithmic scale. Various precipitation ID thresholds for landslide initiation have been derived for different physio-geographical settings and at various spatial scales (e.g. Guzzetti et al, 2007; Guzzetti et al, 2008; Wieczorek and Glade, 2005, and references therein). The global and regional landslide precipitation ID thresholds encompass different types of landslides and a distinct variety of geological and environmental factors, such as lithology, soil depths and land use. The local ID thresholds are restricted more often to relatively homogeneous conditions, and mass movement types.

However, several shortcomings are frequently recognized and discussed. For example, Berti et al (2012) recognized the problem of looking at landslide occurrence and disregarding non-occurrence when applying the ID threshold. They used a Bayesian probability approach to derive the probabilistic transition explicitly taking into account landslide occurrence and non-occurrence. Also the role of hydrology in landslide initiation, although often acknowledged to be of key importance, is usually not included in the statistical precipitation ID threshold approach. Some attempts to more explicitly include hydrology have been proposed, but however they were mainly limited to include measures for antecedent soil moisture content. However, to the authors' knowledge, these studies have never been subject to a more thorough analysis of the specific hydrological information needed for reliable local and regional hazard prediction.

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

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102 2 HYDROMETEOROLOGICAL ANALYSIS OF ID THRESHOLDS

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104 COMPARING ID THRESHOLDS WITH IDF CURVES

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

$$115 \quad I = A^B \quad (1)$$

116

117 with B [-] being the slope of the log-plotted straight line and A [mm/hr] a measure of the rain
 118 intensity of a rain event of unit duration.

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

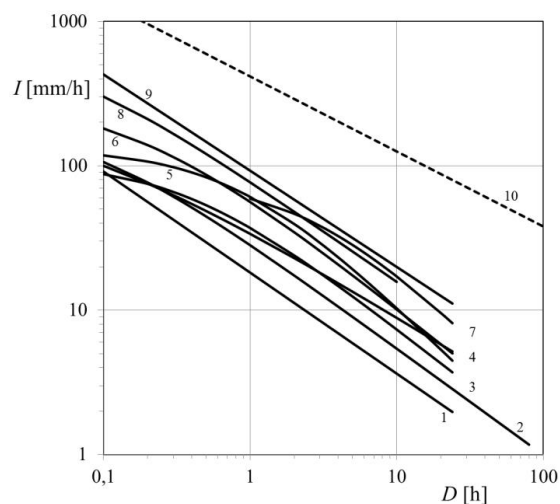


Figure 1. Examples of intensity-duration-frequency curves for 10 years return period (1-9) and curve of the maximum observed precipitations (10). Location and source: 1 Najran region, Saudi Arabia (Elsebaie, 2012); 2 Uccle, Belgium (Van de Vyver, 2015); 3 Naples, Italy (Rossi and Villani, 1993); 4 Los 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).

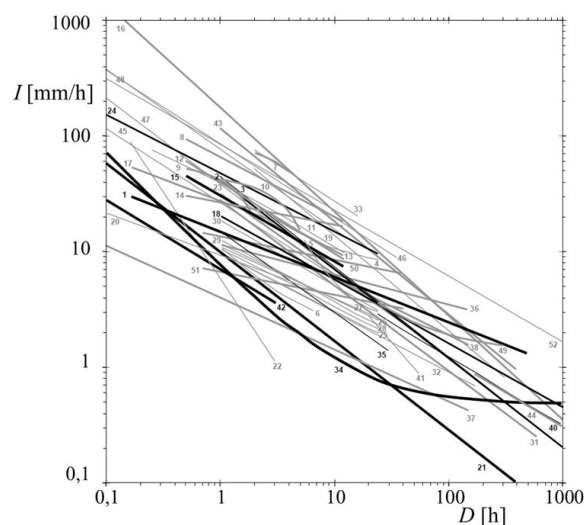


Figure 2. Rainfall intensity-duration (ID) thresholds. Numbers refer to case studies (Guzzetti et al, 2007). Very thick lines are global thresholds; thick lines are regional thresholds and thin lines are local thresholds. Black lines show global thresholds and thresholds determined for regions or areas pertaining to the Central to Eastern European region. Grey lines show thresholds determined for regions or areas not-pertaining to this area.



Figure 2 shows the thresholds that come from a global dataset of landslides, more than 90% of which are shallow landslides and debris flows (Guzzetti et al, 2007). Note that the thresholds are usually obtained as lower envelope of the events resulting in landslide initiation, although also other thresholds definitions exist (e.g. Staley et al., 2013; Peres and Cancelliere, 2016, Ciavolella et al, 2016). Obviously, ID thresholds differ greatly between climate and physiographic regions, especially in absolute values. Therefore, scaled representations have been proposed for the thresholds, such as dividing precipitation intensity by the mean annual precipitation (Guzzetti et al, 2007), in order to better compare the thresholds. However, in our analysis the focus will be on the absolute precipitation ID representation, as it is a convenient way to compare with IDF and for the following discussion. The exponent of most of the reported thresholds for initiation of landslides range between -0.2 and -0.6. By overlaying IDF and ID curves (Figure 3), for landslides triggered by short precipitation events ($D \leq 1$ hr), mostly debris flows and some shallow landslides, the slopes of ID and IDF curves substantially coincide. On the other hand, for longer precipitation durations, ID thresholds have smaller slopes than IDF curves. This means that landslide initiation on the right side of the graph (lower precipitation intensity with longer duration) would occur with rapidly increasing return periods of precipitation events. This is counter-intuitive, as during long-lasting wet periods landslides are usually more frequent. This shows that the method used to derive ID thresholds for landslide initiation based on landslide and precipitation reports leads to troublesome interpretations. Several authors already pointed out that characterizing a storm with its mean intensity, thus neglecting peaks and underestimating actual intensity, affects the estimated probability of landslide occurrence (e.g. D’Odorico et al., 2005; Peres and Cancelliere, 2016), and such an issue is obviously more significant for long storm durations.

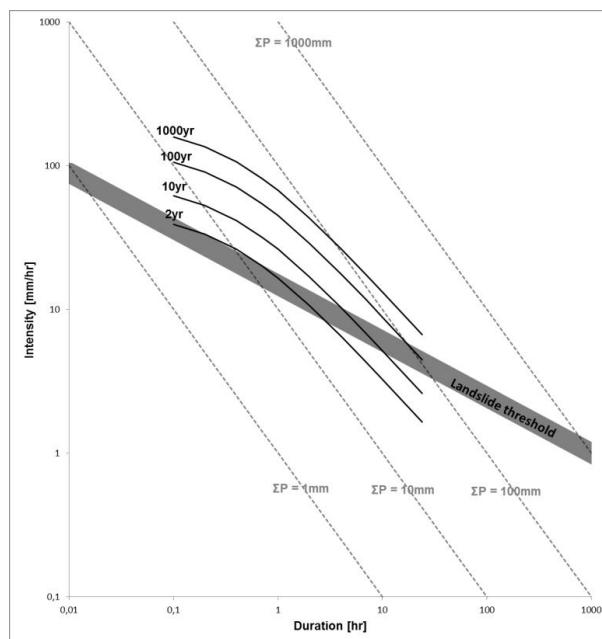


Figure 3. Schematic representation of precipitation IDF curves, isolines of accumulated precipitation and generalised ID threshold for shallow landslides and debris flows (derived from Figure 2).

HYDROLOGICAL INTERPRETATION OF ID THRESHOLDS

The precipitation ID thresholds are “*volumetric*”, i.e. they depict the total, cumulative amount of precipitation. In Figure 3 the global summary of ID thresholds for shallow landslides and debris flows (Guzzetti et al, 2007) is schematically represented by the dark grey area, but, added to it, are isolines of accumulated precipitation volume (1, 10, 100 and 1000 mm). The first observation is that the regional and global landslide thresholds clearly follow a slope different from isolines, meaning that longer duration landslide triggering thresholds require larger water volume. This is understandable if landslide size increases. However, the database consists for the overwhelming majority of shallow landslides and debris flows (Guzzetti et al, 2007). Clearly, this indicates the role of hydrology, or, to be precise, the balance between infiltration, storage and drainage capacity of a slope (Bogaard and Greco, 2015).

If we take a closer look at the volumes needed for landslide initiation, we see a spread in landslide triggers between the roughly <10 mm and >1000 mm of accumulated precipitation, with the vast majority of the empirical precipitation thresholds being reported



188 between 10 and 100 mm of accumulated precipitation. Under ‘normal’ antecedent wetness
189 conditions (that is, soil field capacity), an accumulated precipitation of < 10 mm is generally
190 not capable of triggering a landslide or debris flow. Of course, such an accumulated
191 precipitation volume can trigger a shallow landslide or debris flow in case of nearly saturated
192 antecedent conditions. In this latter case, the reported precipitation event is really the last
193 ‘push’, the so-called trigger (see next section). On the other hand, precipitation volumes over
194 1 meter and/or durations of over 100 or even 1000 hours (> 1 month) are difficult to interpret
195 in terms of average precipitation intensities and triggering thresholds for shallow landslides
196 and debris flows. The point we want to make here is that the current ID concept combines a
197 too wide range of information with different types of hazards (debris flows and landslides
198 relate to different hydrological processes), different temporal meteorological information
199 (from minutes to several days). This makes the use of ID thresholds cumbersome or even
200 misleading.

201 Additionally, ID thresholds have been derived applying physically-based models of
202 infiltration and slope stability evaluation, which allow taking into account soil hydraulic
203 properties, different initial moisture conditions and the boundary conditions through which
204 the slope exchanges water with the surrounding hydrological system (e.g. Terlien, 1998;
205 Rosso et al., 2006; Salciarini et al., 2006; Frattini et al., 2009; Papa et al., 2013; Peres and
206 Cancelliere, 2014). Such physically-based thresholds often do not follow equation 1,
207 generally adopted for ID thresholds. For long precipitation durations, the physically-based ID
208 curves tend to flatten (e.g. Rosso et al., 2006; Salciarini et al., 2006), indicating that landslide
209 initiation thresholds become less sensitive to (average) precipitation intensity, which
210 consequently is not anymore a good explanatory variable for landslide initiation.

211 Interestingly, Frattini et al. (2009) followed an inverse approach and obtained
212 estimates of the probability of the precipitation characteristics leading to shallow landslide
213 initiation by considering also antecedent precipitation. In particular, they showed how the
214 exponent of the IDF curves of their study area (a catchment located on the east side of the
215 Lake Como, in Lombardy, northern Italy) changed from -0.65, for unconditional probability
216 of triggering events, to -0.43, when only events preceded by 300mm fallen in the antecedent
217 four days were considered, thus approaching the slope of the observed ID thresholds.
218 Antecedent precipitation can be seen as an indirect means to take into account the moisture
219 conditions of the soil cover before a triggering event. Therefore, the results of Frattini et al.
220 (2009) can also be interpreted as an indirect confirmation that considering the involved
221 hydrological processes would improve the performance of landslide initiation thresholds.



222 Greco and Bogaard (2016) give an example of the possible inclusion of slope hydrological
 223 processes in the definition of landslide initiation threshold for the case of a slope covered by
 224 loose granular volcanoclastic deposits laying upon a fractured limestone bedrock. The
 225 hydraulic characteristic curves of the volcanic ashes constituting the majority of the soil cover
 226 were known (Damiano et al., 2012; Greco et al., 2013), as well as the moisture state of the
 227 cover before all 78 observed rainfall events (Comegna et al., 2016). Hence, it was possible to
 228 define non-dimensional variables comparing the meteorological triggers with the infiltration
 229 and storage capacity of the soil cover, showing that a non-dimensional hydro-meteorological
 230 threshold performed slightly better than the precipitation ID threshold in separating events
 231 resulting in factors of safety smaller and greater than 1.3.

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233 3 TRIGGER - CAUSE CONCEPT: PROPOSING HYDROMETEOROLOGICAL 234 LANDSLIDE THRESHOLDS

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236 In the strict sense, the precipitation ID threshold is an empirical-statistical threshold drawn to
 237 separate failure and non-failure conditions based on observed landslides and precipitation
 238 records. Precipitation is described in terms of intensity and duration. The main assumption is
 239 that there is an underlying causal relation between the recorded precipitation event and the
 240 landslide occurrence. However, by including durations up to e.g. 1 month, the direct causal
 241 relationship is weak and the method implicitly includes the wetness state of a region. This
 242 limitation has been recognized from the start of using ID thresholds. For several regional
 243 hazard assessment analyses, research groups have extended the ID threshold method by
 244 replacing the duration of a precipitation event on the x-axis with an antecedent precipitation
 245 index (e.g. Crozier and Eyles, 1980; Glade et al, 2000). This, however, leads to limited added
 246 information as still only precipitation information is used. On the contrary, by replacing the x-
 247 axis with a measure for antecedent soil water content, physically relevant information is
 248 added (e.g. Crozier and Eyles, 1980; Wilson 1989; Wilson and Wiczorek, 1995; Crozier,
 249 1999; Glade, 2000; Chirico et al. 2000; Gabet et al, 2004; Godt et al, 2006; Ponziani et al,
 250 2012). Interestingly, by including a water balance of the potentially unstable soil, a statistical
 251 ID threshold evolves conceptually from a plot with one prevalent driver and data source
 252 (precipitation) into a plot containing two predominant drivers with two distinct time-scales:
 253 the antecedent hydrological ‘cause’ and the precipitation ‘trigger’. Besides soil water balance
 254 calculations, different sources of hydrological information can be used to quantify the
 255 hydrological ‘cause’ of landslides. This is a largely unexplored terrain, partly as data



256 availability can be cumbersome and partly because physically-based, (semi-) distributed
 257 modelling was preferred.

258 Concerning the ‘trigger’-axis, there is little debate; it is the short-term last push for a
 259 landslide. The time-scale for local and regional assessment of course depends on the local
 260 situation, but hourly or daily time-scales are the most common. The ‘cause’-axis should
 261 represent the predisposing condition of an area under study. For hydrologically triggered
 262 landslides, it should be related to the wetness state of the area. However, there are several
 263 possible choices of hydrological variables to be plotted along the ‘cause’-axis, such as soil
 264 water content, catchment storage, representative regional groundwater level and similar.

265 As mentioned before, there are –besides the soil moisture storage calculations
 266 described above- various examples of hydrological information added to landslide thresholds.
 267 Hashino and Murota published in 1971 an analysis of landslide triggers in a catchment,
 268 related to debris production, using measured river discharge data to link it to the water
 269 balance of the catchment. They identified that the landslides in their study area took place in
 270 antecedent conditions of catchment storage above average. This is one of the earliest reported
 271 studies we know of explicitly looking at catchment water balance as an important source of
 272 information on the antecedent hydrological condition of an area in relation to landslide
 273 occurrence. Reichenbach et al (1998) made a combined flood and landslide hazard analysis of
 274 the Tiber river catchment using 72 years of historical daily discharge data from different
 275 gauging stations where hydrological parameters were calculated, such as maximum mean
 276 daily discharge, discharge intensity and flood volume and duration. Probability of occurrence
 277 of landslides and floods was based on the ranking of the events. Combining maximum mean
 278 daily discharge and discharge intensity, regional hydrological thresholds for landslide and
 279 flood hazard (individually or combined) could be drawn. Chitu et al (2016) followed a
 280 somewhat similar approach, analyzing the river discharge in several catchments in the
 281 Ialomita Subcarpathians in Romania for landslide events occurred in 2014. The catchments
 282 could be characterized as having low/high storage. Additionally, a calibrated regional rainfall-
 283 runoff model was used for hydrological analysis of landslides in particular catchments.
 284 Detailed analysis of the (modelled) hydrological response indicated that in two catchments
 285 with low permeability the direct runoff was strongly related to landslide occurrence, whereas
 286 it was linked to modeled soil infiltration flux in the more permeable catchment. In some cases,
 287 regional groundwater level could be informative. Bogaard et al (2013) studied the hydro-
 288 meteorological triggering threshold of the re-activating coastal Villerville–Cricqueboeuf
 289 landslide, Normandy, France. In this situation the hinterland of the coastal cliff consists of a



290 well-defined regional groundwater level. Landslide reactivation was seen to take place only
 291 when water level was in the upper, more permeable top layer. The triggering rain event
 292 together with surpassing a certain regional groundwater threshold could explain 3 of the 4 re-
 293 activations. Note that these groundwater levels were not taken in the active landslide area but
 294 several kilometers inland.

295 Recently, further attempts have been made to use river discharge and lumped water
 296 storage in a catchment as a proxy for the predisposing conditions for landslides along its
 297 slopes. Following Hashino and Murota (1971), the basic idea is that when ‘more-than-
 298 average’ water is stored in the catchment, it is more likely that a rainfall event triggers
 299 landslides. The disadvantage of using catchment wide storage is the relatively low spatial
 300 resolution, and the difficulty of having (reliable and homogeneous) discharge time series in
 301 catchments. Moreover, catchment storage assessment needs information on evaporation which
 302 can have significant uncertainties. Of course, such an approach works only if the causes of the
 303 predisposing conditions for landslides are somewhat related to catchment scale hydrological
 304 processes. Ciavolella et al. (2016) defined a cause-trigger hydro-meteorological threshold in
 305 the catchment of river Scoltenna, in Emilia Romagna (Italy), linking catchment storage and
 306 event rainfall intensity, and compared its performance with that of a statistical ID
 307 precipitation threshold. The two thresholds performed similarly, with the hydro-
 308 meteorological thresholds resulting more accurate in the identification of landslides, but
 309 giving a somewhat larger number of false positives.

310 The above examples indicate that considering hydrological causes could be useful for
 311 a better identification of landslide initiation, but, at the same time, they show that the correct
 312 identification of the hydrological processes involved in the establishment of the predisposing
 313 conditions for landslides in a considered area is mandatory for choosing the most informative
 314 hydrological variable to be plotted along the x-‘cause’-axis.

315

316 4 CONCLUDING REMARKS AND OUTLOOK

317

318 The intrinsic limitation of precipitation ID thresholds for the identification of landslide
 319 initiation conditions has been pointed out long since. Indeed, such thresholds neglect the role
 320 of the hydrological processes occurring along slopes, which lead to the predisposing
 321 conditions for failure (causes), and focus predominately on the characteristics of the last
 322 rainfall event leading to slope failure (triggers). As a consequence, the predictive value of the
 323 ID thresholds is often low, even when they refer to small areas. We argue that the threshold



324 values for rainfall intensity of short and long duration (the far left and right side of the graphs)
325 have limited physical meaning and, consequently, that the use of precipitation ID thresholds
326 can lead to misleading interpretation of initiation conditions, as important antecedent
327 conditions and rainfall intensity variations are not taken into account. For this reason, we here
328 advocate to be very careful in uncritically using the precipitation ID thresholds as kind of
329 regional characteristic of (shallow) landslide occurrence.

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

340 We propose to use the cause-trigger concept for defining regional landslide initiation
341 thresholds. This, we agree, is challenging, but, in our opinion, not impossible. Looking at the
342 discussed examples, it becomes clear that the choice of the most informative hydrological
343 variable to be used as a proxy for landslide predisposing conditions strictly depends on site-
344 specific geomorphological characteristics, and that accurate analysis of the boundaries
345 through which the potentially unstable area exchanges water with the surrounding
346 hydrological systems is mandatory. In other words, for the assessment of landslide
347 predisposing conditions, the water balance of the slope should be assessed, but getting the
348 information about the involved hydrological processes (e.g. evaporation, runoff, groundwater
349 recharge) at the required spatial-temporal resolution is often still a challenge, although remote
350 sensing could help increasing the reliability of catchment water storage estimates.

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