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Abstract	Agricultural landscapes cover a significant part of the Earth. In floodplains, we can find large areas dedicated to intensive agriculture. However, also on hills and mountains, agricultural activity can be relevant from the socio-economic point of view. Nowadays, such areas are increasingly under threat because of global environmental changes. Widespread growing rainfall aggressiveness due to climate change, in addition to land abandonment, lack of structural maintenance, and in some cases unsuitable agronomic practices are exposing steep-slope agricultural landscapes to increased hazard of landslides. A	

suitable hazard assessment and zonation of these phenomena would help better management of such agricultural landscapes. The purpose of this article is to provide an overview of this relevant problem focusing on (i) the contribution of remote sensing technologies (e.g., LiDAR and UAV photogrammetry) in mapping the investigated processes, and (ii) discussing advances and limitations of susceptibility modelling.

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Keywords

Landslide - Remote sensing - Modelling - Landscape - Agriculture

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# Landslides in Steep-Slope Agricultural Landscapes

Paolo Tarolli, Anton Pijl, and Sara Cucchiaro

## Abstract

Agricultural landscapes cover a significant part of the Earth. In floodplains, we can find large areas dedicated to intensive agriculture. However, also on hills and mountains, agricultural activity can be relevant from the socio-economic point of view. Nowadays, such areas are increasingly under threat because of global environmental changes. Widespread growing rainfall aggressiveness due to climate change, in addition to land abandonment, lack of structural maintenance, and in some cases unsuitable agronomic practices are exposing steep-slope agricultural landscapes to increased hazard of landslides. A suitable hazard assessment and zonation of these phenomena would help better management of such agricultural landscapes. The purpose of this article is to provide an overview of this relevant problem focusing on (i) the contribution of remote sensing technologies (e.g., LiDAR and UAV photogrammetry) in mapping the investigated processes, and (ii) discussing advances and limitations of susceptibility modelling.

## Keywords

Landslide • Remote sensing • Modelling • Landscape • Agriculture

## Background

Agricultural land use is responsible for an unprecedented transformation of natural environments worldwide, with vast and long-term impacts on geomorphology and soil properties (Bartman et al. 2012). Cultivated areas typically

involve reduced soil cover and cohesion, which particularly in steep-slope environments greatly affect soil erosion and slope instability (Koulouri and Giourga 2007; Prosdociami et al. 2016; Tarolli and Straffellini 2020). Several researchers around the world have studied the increased landslide susceptibility of cultivated hillslopes, and several factors related to agricultural practices (i.e. not considering climate or slope steepness) have been discussed. Agricultural transformation affects soil stability due to the removal of permanent deep-rooted vegetative cover (Perotto-Baldivieso et al. 2004), while the natural soil structure is affected due to land levelling (DeGraff and Canuti 1988; Ramos et al. 2007). In addition, an unsuitable or degraded terracing or drainage system can further aggravate landslide hazard (Tarolli et al. 2014). This is often related to the lack of maintenance as a result of land abandonment and loss of labour, as widely reported for Mediterranean (Arnáez et al. 2017; Cevasco et al. 2014; Tarolli et al. 2014) and Asian steep-slope agricultural areas (Gerrard and Gardner 2002; Raj Khanal and Watanabe 2006). Other factors include the cultivation and subsequent reactivation of dormant landslides (Sugawara 2013), and the construction of agricultural roads for machinery (Tarolli et al. 2015). The latter is able to divert runoff and create concentrated patterns of water flow, which are often related to the initiation of landslides (as illustrated in Fig. 1).

The high landslide hazard in agricultural areas can have considerable impact on production and human safety. Nonetheless, reliable inventories are missing for many marginalised steep rural areas around the world. Modern developments in remote sensing, computer technologies and models may help contributing to this. In this work, we discuss the opportunities and challenges of remote sensing techniques, digital terrain analysis and landslide modelling, based on literature and few original examples.

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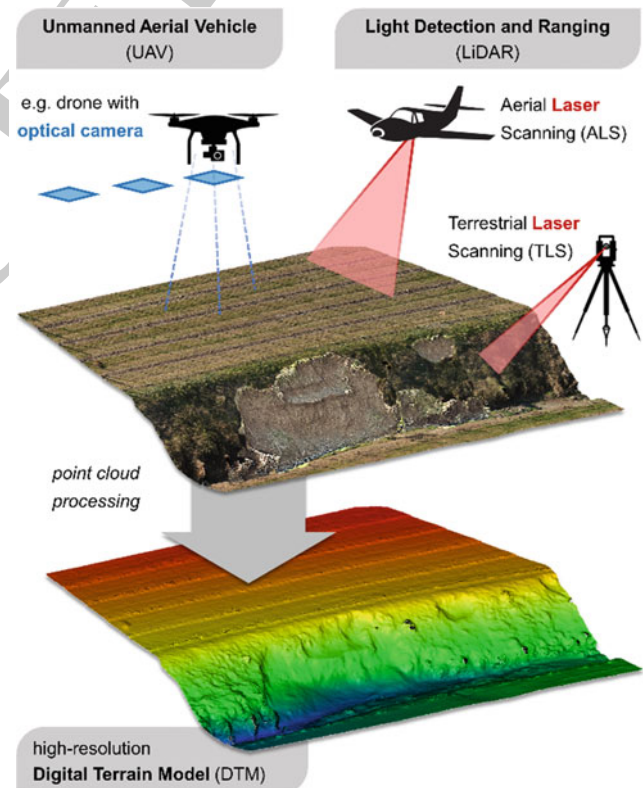
**Fig. 1** Landslides that occurred after an intense rainfall event in a typical steep-slope vineyard terraced landscape. Landslide crowns are highlighted with a white dashed line (Photographs by P. Tarolli)



## Remote Sensing

### Techniques

Landslide inventory maps are generally limited in terms of spatial coverage and time period (Guzzetti et al. 2012), which can partly be attributed to the intensive mapping methods used in the past (Galli et al. 2008). However, there is a strong potential to address this gap by the use of modern remote sensing techniques, which allow more accurate topographic analysis by use of faster and cheaper surveys (Tarolli 2014). The most recent platform that proved highly successful is the use of an Unmanned Aerial Vehicle (UAV, sometimes referred to as UAS or RPAS) and the parallel development of Structure from Motion (SfM) photogrammetry technique (Giordan et al. 2018), which allows high-accuracy surveying by low-cost applications of a simple drone mounted with a non-metric camera (Fig. 2). This platform rapidly gained popularity for mapping topographic features, as it is flexible to deploy in varying conditions (including difficult-to-access sites) and is able to capture complex geomorphologic features (Eltner et al. 2016; Cucchiniaro et al. 2018). The typical coverage of light-weight UAVs is in the order of tens of hectares and is optimal for agricultural conditions (Colomina and Molina 2014; Pijl et al. 2020). Indeed, specific examples of UAV-based analysis of agricultural slope failure can be found, e.g. for monitoring mass-movements (Lucieer et al. 2014; Turner et al. 2015), or for the detection and modelling of terrace failures (Pijl et al. 2019).



**Fig. 2** Schematic illustration of the creation of a high-resolution Digital Terrain Model (DTM) by remote sensing techniques of Unmanned Aerial Vehicles (UAVs) or Light Detection and Ranging (LiDAR) instruments (Sample dataset from original data by the authors)



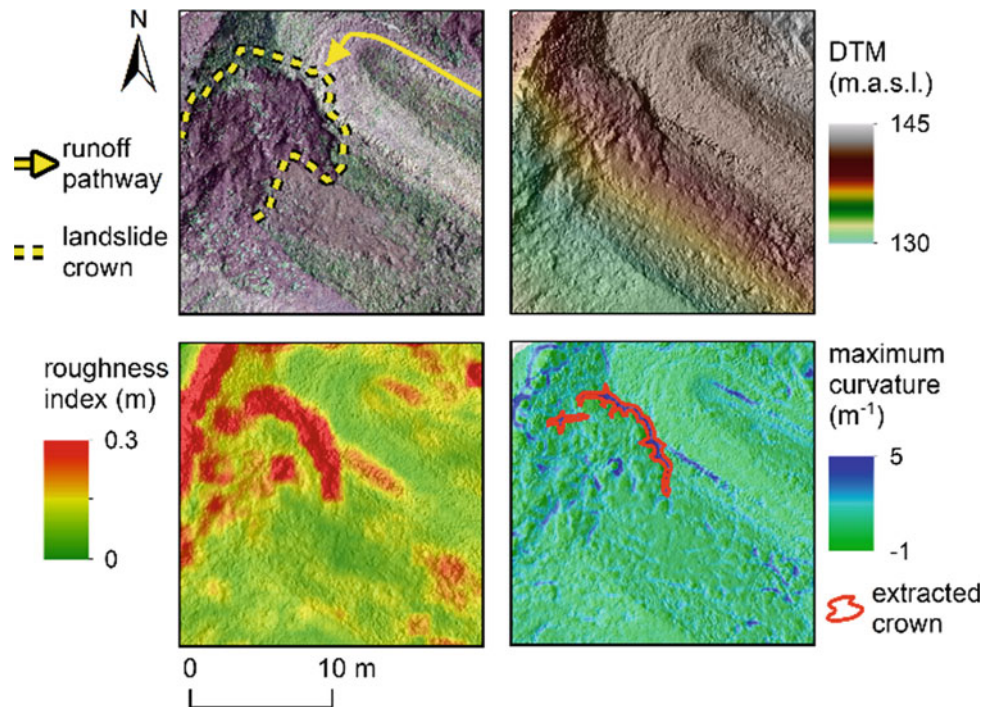
Another modern remote sensing technique that is commonly used for high-resolution topographic mapping is Light Detection and Ranging (LiDAR) from a terrestrial or aerial platform (Fig. 2). It is probably one of the most popular technique in landslide research, by providing highly-detailed topographic information for the detection, monitoring or modelling of slope failure (Jaboyedoff et al. 2012; Scaioni et al. 2014). Although LiDAR involves considerably higher deployment costs than UAV-based surveys (Guzzetti et al. 2012), it offers a major advantage over the latter due to its ability to retrieve multi-layered return signals and resulting ground detection in densely vegetated areas (Tarolli 2014). As such, it is probably the most established platform for geomorphologic research, and countless examples can be found of LiDAR-based research of agricultural landslides, e.g. in Italian terraced landscapes (Preti et al. 2018a; Tarolli et al. 2014, 2015; Giordan et al. 2017; Brandolini et al. 2018).

### Digital Terrain Analysis

The improving spatial accuracy of topographic data (e.g. from UAV or LiDAR source) also allows higher accuracy in the detection of landslides and their geomorphology. Basic 2-D digital terrain analysis can be used to support visual detection, while facilitating more advanced purposes such as volume estimation, multi-temporal monitoring and susceptibility modelling (as discussed in the following chapter). Several authors have tested the use of terrain derivatives such as slope, surface roughness or curvature for the automatic delineation of geomorphologic features using a statistical threshold (McKean and Roering 2004; Booth et al. 2009; Tarolli et al. 2012). In this work, we illustrate this concept using an original topographic data sample, consisting of a road-induced landslide (Fig. 3). The shaded 0.05-m DTM clearly shows how the landslide crown is captured, appearing as a dark edge at the road side. This crown presents a disruption with respect to the surrounding geomorphology, as can be identified using the surface roughness index (Fig. 3), here calculated according to Cavalli et al. (2008) using a 31-cell kernel. Similarly, high values of maximum landform curvature indicate convex terrain elements (according to Evans 1979 using a 31-cell kernel); and the statistic threshold approach by Sofia et al. (2014) has been used to automatically extract the landslide crown (Fig. 3). This approach could rapidly be applied over large extents, hence becoming a powerful tool for automatic geomorphologic feature extraction.

### Modelling

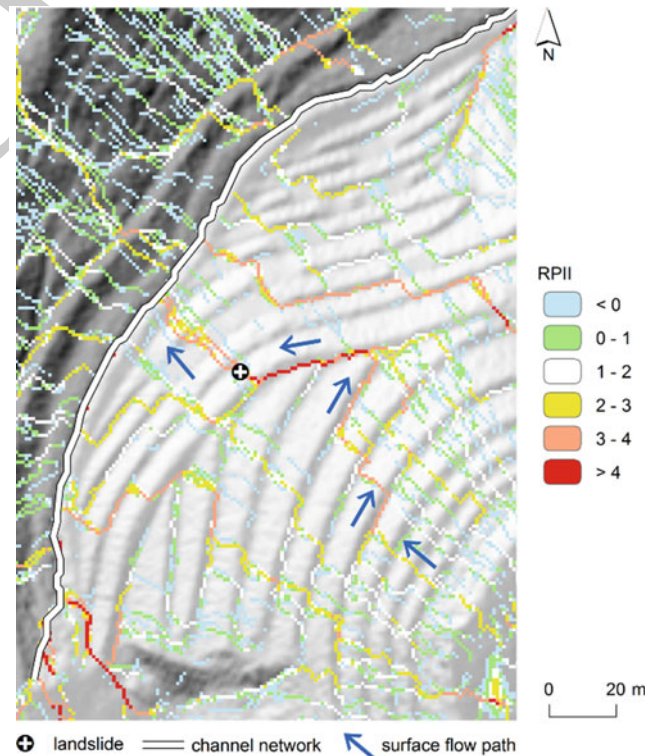
For estimating the susceptibility, or spatial probability of slope failures occurrence, numerous models are available (Guzzetti et al. 2005), but the number of publications on landslide hazard assessment in steep agricultural slopes is still rather modest. In this anthropic context, further variables must be considered than the classic ones used to better describe the landslide triggering processes. Here, slope failure hazard is not only dependent on meteorological events and geotechnical land attributes, but also agricultural practices and human activities can strongly affect the occurrence of instability phenomena (Shrestha et al. 2004). For example, Perotto-Baldiviezo et al. (2004) developed and tested a spatial model for predicting the spatial distribution of landslide hazard in steep areas of Honduras considering four model variables as slope, aspect, stream proximity and land cover type. Their results highlighted how. As the slope increased, the percentage of land affected by landslides, increased sharply on cropland when the soils were saturated. This indicated that agricultural activity and the associated removal of deep-rooted permanent vegetation increased the landslide hazard on steep sites. Jaiswal et al. (2010) also proposed a probabilistic landslide model to quantify hazard of first-time slope failure on natural slopes and tea plantations, underlining how agricultural and constructional activities make an area more susceptible to landslides. At the local scale, numerical simulations must consider the presence of widespread agricultural practices such as terraces that heavily influence the hydrological processes (Gallart et al. 1994; Preti et al. 2018a) and can trigger superficial mass-movements when they collapse. Camera et al. (2014) analyzed the processes that can lead to failure. They used a numerical modelling of groundwater movement and related stability analysis, to provide pore water pressure distributions, which are generated by different rainfall amounts, as parameters for a stress-strain analysis that can directly determine the influence of various rainfall parameters on dry-stone wall stability. Penna et al. (2014) evaluated the predictive power of the quasi-dynamic shallow landslide model QD-SLaM (Tarolli et al. 2008) to simulate shallow landslide locations in a small-scale steep-slope watershed. The study area represented a typical anthropogenic Mediterranean landscape, with terraces (partly abandoned), roads and a village. The applied landslide model did not incorporate the description of road-related or terrace-related failures, thus highlighting its limits in the correct interpretation of hydrological processes in an anthropogenic context. The model predictive power was shown to be



**Fig. 3** Example of a landslide induced by a runoff pathway on a curving agricultural road (top-left, yellow lines). Additionally, illustrated are the shaded high-resolution DTM derived by UAV-SfM workflow (top-right), and two terrain derivatives: the roughness index (bottom-left) and the maximum landform curvature (bottom-right). The latter shows the automatically extracted landslide crown using a threshold approach (red outline). (Sample topographic dataset with 0.05-m spatial resolution by the authors)

DTM-resolution dependent. The use of a coarser resolution had a smoothing effect on terrain attributes, and therefore on predictive model performance. The authors concluded that to realize the full potential of high-resolution topography thus including anthropogenic geomorphic features, more extensive work is needed to specifically identify the extent of the artificial structures and their impact on shallow landsliding processes. Tarolli et al. (2015) used LiDAR elevation data for a detailed hydro-geomorphological analysis of terraced vineyards. The geomorphic Relative Path Impact Index (RPII) was tested in two vineyards to identify terrace-induced and road-induced erosion. Using such an index, the authors then simulated different scenarios of soil conservation measures, establishing the optimal solution to reduce erosion. The results highlighted the effectiveness of high-resolution topography in the analysis of erosion at the local scale of terraced vineyards when surface water flow is the main factor triggering the instabilities. An example of the predictive power of RPII in the recognition of potential slope failure in a steep-slope terraced site is shown in Fig. 4, where we considered the case study of Fig. 1.

Preti et al. (2018b) proposed a more specific model able to describe hydrological processes in terraced landscapes. In detail, they analyzed the destabilizing pressures acting on the retaining dry-stone walls in the most critical portion of each terrace. The results showed good capability of the model to



**Fig. 4** Relative Path Impact Index (RPII) calculated for the case study shown in Fig. 1 using a 1-m LiDAR-derived DTM. Surface flow modifications induced by an agricultural road and consequent landslide (warm colours) match well with the location of the observed landslide (plus symbol), thus confirming the predictive power of the index for the detection of slope failure in this context



226 predict the distribution and intensity of stress on the  
227 dry-stone wall over time and space. This stress was related to  
228 the combined earth pressure and hydrostatic pressure (water  
229 accumulation), without the occurrence of soil saturation.  
230 A better understanding of the main hydrological processes  
231 that govern surface and subsurface water flow pathways and  
232 that are responsible for terrace failure is essential for  
233 appropriate water resource management and rural landscape  
234 maintenance in terraced areas (Preti et al. 2018a). Such  
235 insights could support landowners and land planners in  
236 managing these complex and fragile environments.

### 237 Final Remarks

238  
239 Climate change (e.g., the increase of rainfall intensity) and  
240 changing societal trends (e.g., land abandonment) are  
241 aggravating land degradation in agricultural landscapes,  
242 resulting in increased mass movements that should not be  
243 neglected. Not only does this affect agricultural production,  
244 it also poses a risk to local communities of people. Several  
245 approaches for the spatial analysis of such processes are  
246 available, however high-resolution topography derived by  
247 modern remote sensing techniques (e.g., SfM photogram-  
248 metry using UAV images, or airborne or terrestrial LiDAR  
249 data) is highly recommended for the detection of the char-  
250 acteristic local-scale geomorphic features usually visible  
251 with sub-meter DTM grid cell size. In addition, it is also  
252 recommended to improve the understanding of the main  
253 hydrological processes that govern surface and subsurface  
254 water flow pathways. Existing modelling approaches seem  
255 to be not optimal for a satisfactory understanding of such  
256 processes. Indeed, to meet the full potential of  
257 high-resolution topography, they should be designed to  
258 include the anthropogenic geomorphic features and model  
259 their impact on the physical processes or vice versa. More  
260 recently, few authors developed some advances along this  
261 line. This will absolutely be a future challenge for scientific  
262 studies in this field. Novel insights from research could  
263 support land owners and land planners in managing these  
264 complex and fragile landscapes, in order to preserve cultural  
265 heritage, ecosystem services, and food safety while main-  
266 taining the economic and environmental sustainability.

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418 *Remote Sens* 7:1736–1757. <https://doi.org/10.3390/rs70201736> 467

Author Proof





## Author Query Form

Book ID : 476715\_1\_En

Chapter No : 46

Please ensure you fill out your response to the queries raised below and return this form along with your corrections.

Dear Author,

During the process of typesetting your chapter, the following queries have arisen. Please check your typeset proof carefully against the queries listed below and mark the necessary changes either directly on the proof/online grid or in the 'Author's response' area provided below

Query Refs.	Details Required	Author's Response
AQ1	Kindly note that the cross citation of reference 'Koulouri et al. (2007), Arnáez et al. (2011)' has been changed to 'Koulouri and Giourga (2007), Arnáez et al. (2017)' so that this citation matches the list. Please check and confirm.	

# MARKED PROOF

## Please correct and return this set

Please use the proof correction marks shown below for all alterations and corrections. If you wish to return your proof by fax you should ensure that all amendments are written clearly in dark ink and are made well within the page margins.

<i>Instruction to printer</i>	<i>Textual mark</i>	<i>Marginal mark</i>
Leave unchanged	... under matter to remain	Ⓟ
Insert in text the matter indicated in the margin	∧	New matter followed by ∧ or ∧ <sup>Ⓢ</sup>
Delete	/ through single character, rule or underline or ┌───┐ through all characters to be deleted	Ⓞ or Ⓞ <sup>Ⓢ</sup>
Substitute character or substitute part of one or more word(s)	/ through letter or ┌───┐ through characters	new character / or new characters /
Change to italics	— under matter to be changed	↵
Change to capitals	≡ under matter to be changed	≡
Change to small capitals	≡ under matter to be changed	≡
Change to bold type	~ under matter to be changed	~
Change to bold italic	≈ under matter to be changed	≈
Change to lower case	Encircle matter to be changed	≡
Change italic to upright type	(As above)	⊕
Change bold to non-bold type	(As above)	⊖
Insert 'superior' character	/ through character or ∧ where required	Υ or Υ under character e.g. Υ or Υ
Insert 'inferior' character	(As above)	∧ over character e.g. ∧
Insert full stop	(As above)	⊙
Insert comma	(As above)	,
Insert single quotation marks	(As above)	Ƴ or ƴ and/or ƶ or Ʒ
Insert double quotation marks	(As above)	ƶ or Ʒ and/or Ʒ or ƶ
Insert hyphen	(As above)	⊥
Start new paragraph	┌	┌
No new paragraph	┐	┐
Transpose	└┐	└┐
Close up	linking ○ characters	⸸
Insert or substitute space between characters or words	/ through character or ∧ where required	⸶
Reduce space between characters or words		⸵