1. Introduction

The fast proliferation of new satellite, aerial and terrestrial remote sensing techniques has undoubtedly provided new technological and scientific opportunities to society during the last few decades. Driven by this impressive technological progress together with the development of innovative data treatment techniques and the progressive adaptation of these workflows by geoscientists, the way we see, understand, and model the Earth’s surface processes has been greatly enhanced [1]. Three-dimensional survey techniques, such as LiDAR (or laser scanning) and photogrammetry, which can be mounted over fixed or mobile platforms, including Unmanned Aerial Vehicles (UAV), airborne, helicopter, vessel and others ways of transportation, are completely changing our perception of the surrounding environment [2]. Classical photogrammetric workflows are being progressively replaced by Structure-from-Motion (SfM) workflows imported from computer vision disciplines, allowing much higher versatility and usability by non-experts [3,4]. The output of these techniques—3D point clouds, i.e., high resolution three-dimensional sets of data points defined by X, Y and Z coordinates—are progressively being used in different branches of Earth sciences, including geomorphology, engineering geology, stratigraphy, tectonics, mining, etc. The use of more accurate and denser 3D information terrain surfaces is changing classical investigative approaches for natural phenomena at different spatial and temporal scales, including the mapping of geomorphic features [5], characterization of geological outcrop layers [6,7], investigation of topographic changes [8], identification of structural patterns, study of depositional forms [9], analysis of mass movements [10], and rock slope processes [11]. The investigation of a series of natural processes which may affect human activities and resulting in risk to society using three-dimensional techniques has also attracted special attention by researchers, technicians and policymakers. Several avenues of research have been opened with the development of new automatic or semi-automatic techniques for more advanced observation, monitoring, modelling and prediction of natural phenomena, considerably impacting our society.

This Special Issue contains a collection of articles discussing the use of three-dimensional remote sensing techniques for the investigation diverse geo-hazards. We received a total of 13 manuscripts for consideration, which were thoughtfully reviewed by external experts in their respective fields. In addition, the Guest Editors of this Special Issue carried out a complementary editorial review. As a result, five manuscripts were finally accepted. These five publications reflect a wide spectrum of different sensors covering topics of investigation techniques and analysis methods that can be used for characterizing and monitoring geo-hazards. Together with well-established LiDAR systems, fast evolving image processing techniques were also utilized in our Special Issue. Several submissions were related to the development of specific methods for data treatment, examples including the pre-processing of High Dynamic Range (HDR) imagery, the automatic extraction of rockfalls from multi-temporal LiDAR datasets and the treatment of massive amounts of geometrical data for carrying...
out efficient 4D monitoring. Several other contributions deal with the investigation of a given physical phenomena including slow-moving lava flows coming from recent volcanic eruptions [12], shear displacements due to historical earthquakes [13] and mass movements both on mountainous areas [14] and rocky cliffs [15,16]. A short review of each topic of investigation addressed in this Special Issue, including an overall context and main findings, is given below.

2. Use of 3D Remote Sensing Techniques by Geoscientists: A Bibliometric Analysis

The use of 3D point clouds for the investigation of Earth surface processes is now starting to be a well-established discipline. To show this, we carried out a bibliometric analysis on the occurrence of the terms “Structure from Motion” and “LiDAR” or “laser scan*” in Web of Sciences bibliographical database restricted to its Geosciences segment (Figure 1), in line with [17]. While it can be observed that the use of SFM term on geoscientific literature is very recent, the growth of the number of references related to laser scanning is exponential-like since the 90’s. However and according to [18], the global scientific output doubles every nine years, as observed in the Geosciences segment of the Web of Sciences database (12,540 records for 2000 and 21,445 records for 2010). That leads to the question whether the LiDAR literature is growing significantly faster than the rest of the geoscientific literature. To answer this question, the number of references related to laser scanning has been divided by the total number of references (both numbers restricted to the Geosciences segment of Web of Sciences). Figure 1 shows that the bibliographical impact of laser scanning studies is growing much faster the rest of geoscientific literature, and that very soon, if not already achieved, 2% of the geoscientific literature will be directly linked to LiDAR technologies.

Figure 1. Number of publications in the database Web of Sciences containing the terms “LiDAR” or “laser scan*” in the field TOPICS between 1970 and 2014. Analysis made on January 2016 and restricted to the “Geosciences-multidisciplinary” segment of Web of Science.

3. Summary of the Special Issue Contributions

3.1. Studying Volcanic Features and Active Lava Flows

Using TLS for the investigation of volcanic processes is a challenging topic. Up to now, TLS has mainly been used for the investigation of dormant or quiescent volcanos, examples include the topographic modelling of volcanic cones [19], the 3D reconstruction of volcanic facies [20], the lithological discrimination based on backscattered intensity signals [21], the investigation of
geomorphic changes [22] and other engineering geological applications [23]. The near-real-time investigation of dynamic surface processes such as lava flows on active volcanoes constitute a great challenge nowadays. The article by Slatcher et al. [12] published in this Special Issue shows an impressive series of optical and laser measurements at Mount Etna during a volcanic eruption in summer 2012. TLS measurements were acquired at a high temporal frequency (1 scan every ten minutes), and were complemented by even more frequent time-lapse images acquired each 30 s. The thus-acquired point clouds were interpolated on a high resolution DEM, which were later analysed using conventional DEM subtraction techniques. In addition, instantaneous surface velocities were calculated through feature tracking using both user-supervised image analysis and automatic Particle Image Velocimetry (PIV) technique [24]. Both volumetric fluxes and surface velocities were used for deriving lava rheology, discharge and average cone growth rate. Authors demonstrated how bi-dimensional methods for estimating discharge measurements from satellite images likely result in overestimates, supporting the great potential on the use of ground-based 3D remote sensing techniques in hazardous volcanic areas.

3.2. Monitoring Small Surface Changes Using Noise-Filtering Techniques

Up until very recently, the ability to detect terrain changes using TLS was mainly limited by the achievable accuracy (and resolution) of the input point clouds. While the comparison of multi-temporal datasets using standard point-to-point (or point-to-mesh) procedures does not allow the exploitation of the high data redundancy available on the input data, more elaborated approaches allow considerable reduction in the Level of Detection (LoD) [25–27]. For instance, authors in [25] showed that the characteristic noise locally follows a Gaussian population, allowing to correct the scattering of the TLS differences around the “correct” mean through filtering the point clouds differences in the spatial dimension. Few years later, authors in [26] developed the M3C2 algorithm based on the measurement of the mean surface change along the normal direction. This algorithm based on position averaging has gained a wide acceptance on the geomorphological community during the last years.

The article published in this Special Issue by Kromer et al. [15] goes one step further and takes advantage of point redundancy in space and time (4D) for denoising the point cloud by averaging the differences on several multi-temporal point clouds. These authors also utilise a slope-dependent vector of comparison in line with the M3C2 algorithm [26]. The proposed algorithm allow the significant reduction in the errors and a considerable improvement in the level of detection compared to other well-established monitoring techniques. When its application is possible, it yields feasible investigating sub-mm changes such as diffuse erosion or pre-failure deformation on complex rock-slopes. Due to the 4D nature of the algorithm, both spatial and temporal averaging is adaptable. This computationally efficient algorithm will undoubtedly change current approaches for continuous TLS scanning.

3.3. Rock Slope Monitoring and Automatic Rockfall Extraction

Extracting rockfall magnitudes, frequencies and locations from high resolution, multi-temporal 3D datasets has been a lively research subject during the last decade. Examples include the investigation of rocky cliffs shaped by sea erosion in the UK [28,29] and France [30], the investigation of fragmental rockfalls on mountainous areas [31–34] and the investigation of geohazards along transportation corridors throughout the world [35,36] etc. From a methodological point of view, most of the above mentioned work utilizes a 2.5D interpolative approach for DEM subtraction and for extracting the failure volumes by comparing multi-temporal data. Although most of the procedures recently developed for investigating landslides using 3D point clouds were based on TLS data, several procedures can also be used from photogrammetrically derived point clouds; nevertheless, the number of scientific articles using classical or modern photogrammetric workflows for investigating mass movements in rock slopes is still scarce.

The work presented in this Special Issue by Olsen et al. [16] utilized a 2.5D connected component algorithm for isolating individual failure clusters and for calculating the volume of each individual
rockfall. Compared to other processing techniques using the original 3D point clouds (e.g., noise reduction and feature extraction, clustering of individual failures, automatic volume calculation have already been proposed [32,37]), Olsen and colleagues proposed an original technique for filling data gaps due to topographic obstructions on the interpolated mesh. Both the effects of hole filling and model resolution on their analysis were evaluated, authors observing that model resolution plays a key role in the capability to distinguish smaller rockfall events, which ultimately may have an important influence on the Magnitude-Frequency (MF) relationships. Using TLS to investigate rockslope hazards is starting to be a well-established discipline nowadays [10].

3.4. Detecting Earthquake Signals from Topographic Data

Quantifying the seismic hazard on a tectonically active area is not a difficult task always straightforward [38]; A well-established method for assessing the seismic hazard consists on the analysis of channel offsets and other morphotectonic features on the geomorphological record [39]. More recently, the use of high-resolution three-dimensional datasets is providing new perspectives for the quantification of these low amplitude features along both fast-moving and slow slip-rate faults [40–42]. Unfortunately, there was no standard methodologies to rate the quality of the lateral offsets up until very recently and a certain ambiguity could have affected these studies, as was pointed out by Ferrater et al. [13], in this Special Issue. These authors present a novel methodology for scoring lateral offsets: the Cumulative Offset Probability Density (COPD). They applied this novel method to a slow-moving, strike-slip fault located on the SE of the Iberian Peninsula: the Alhama de Murcia fault. The authors utilized a series of independent variables in an effort to evaluate the quality of their analyses. Their detailed observations can be consulted as kmz files in the supplementary material section of their paper. Interestingly, it appears that slip rates varied on different segments of the fault; as clearly stated by the authors, which may also be due to a lack of specific data affecting their results.

3.5. Structure-from-Motion and Image Pre-Processing Techniques

The emergence of SFM in recent years has revolutionized topographic surveys in geomorphological studies, making data collection and processing techniques available to non-specialists [4]. This trend is even more pronounced in challenging areas where the use of UAV systems is able to overtake several issues affecting conventional photogrammetric techniques, including difficult accessibility, variable distance-to-target and presence of occlusions due to topographic complexities. The article by Gómez-Gutierrez et al. [14] published in this Special Issue evaluated the quality of SFM point cloud in a high-mountain rock glacier in the Pyrenees. Authors assessed the quality on the point cloud generation using both point to point and M3C2 strategies of comparison using TLS point cloud as a benchmark. They tested point cloud generation with several SFM software packages, source imagery, dates of acquisition and presence/absence of snow cover. A LoD of about 30 cm was obtained on such a challenging environment, which is equivalent to some other recent examples of SFM in challenging situations [3,4,43]. Although the accuracy of point cloud reconstructions using different cameras and software packages was evaluated before [44,45], the authors of this publication also tested whether or not the use of High Dynamic Range (HDR) images might produce more accurate results than the use of conventional exposition. Finally, they showed that no significant improvements were obtained using the alternative HDR exposition, which is also a useful piece of experience.

4. Current Challenges in the Use of 3D Point Clouds in GeoHazards

The interest on the characterization and monitoring of geohazards both using LiDAR and photogrammetric techniques is continuously growing as discussed before (Figure 1). SFM workflows have evolved very rapidly during the past ten years, being the potential of SFM for geological studies explored by a growing number of Geoscientists nowadays [4,46]. It is very likely that these workflows will continue to evolve very rapidly during the forthcoming years, and that SFM techniques will turn
into a conventional procedure during the next decade, which will certainly produce a series of exciting research findings and new applications for Natural Hazard assessment.

In spite of recent technological advances, a great challenge remains in the development of new 3D computational procedures for improving characterization, monitoring and modelling of geohazards. Further investigation in the development of new algorithms for process automatization including feature extraction, massive data acquisition, direct georeferencing and real-time monitoring is needed. Another aspect that will grow during the forthcoming years include an integration of very high quality 3D data with current physical models.

Most importantly, we noticed that although high quality measurements can be acquired nowadays with little effort, publishing relevant findings (e.g., presenting new workflows for data treatment, or showing original strategies or techniques for data acquisition, or discussing innovative results helping the understanding of a given physical phenomena) from high quality data is not always straightforward. Using well-established procedures to answer an exciting research question validates and reinforces previous innovations, but on the contrary, repeating the same procedures that have already been used elsewhere should not be further rewarded by editors or external reviewers. Scientific efforts should be placed on real innovation, including the development of new processing techniques or the extraction of some original information from these measurements that helps to a better process understanding.

Lastly, the scientific community has an increasing interest in open knowledge, including open-access publications and open-data. While open publishing seems to be more and more requested by funding agencies (and articles published in *Remote Sensing* can be openly accessed world-wide), other aspects of open knowledge are also very important for the scientific community, including the creation (or maintaining) of open databases containing freely available data and code (or pseudo-code) of already published algorithms. Interestingly, two publications in this Special Issue provide supplementary files: RAW 3D point clouds were provided by [15] and a series of digitized features from which lateral offsets were calculated was provided by [13]. Needless to say, releasing open data in public repositories such as OpenTopography and other public repositories [47–49] considerably helps for validating own results and for allowing geoscientists to advance further in several unexpected directions.

5. Forecasting Future Trends Analysing Bibliographic Data

After having examined the state-of-the-art on the use of 3D remote sensing techniques in the geosciences, one may ask how this situation will evolve during the forthcoming decades. In 2010, authors in [17] analyzed the increasing importance of laser scanning technologies in the geoscientific community, by measuring the number of occurrences of the terms “LiDAR” or “laser scan*” in the keywords of the Georef bibliographical database. They illustrated the fast growing impact of laser scanning in geosciences research based on the 2009 version of the database. They also forecasted future developments of this technique, stating that by 2016 laser scanning should become a routine method and that its popularity will reach a ceiling (Figure 2). This forecast was based on a gentle reduction of the increasing number of references observed in the catalogue for the years 2005–2008, which was fitted with a logistic function. The same analysis was performed in 2014 and again in 2016, showing that this forecast was wrong. Indeed, (Figure 2) shows that the number of references to laser scanning in the geoscientific literature has continued to increase almost exponentially. The apparent slowing down observed in previous analysis [17] was actually due to the fact that it requires three to four years to complete the catalogue of bibliographical references in the Georef database. Using the catalogue of January of 2016 of Georef, and considering that it is almost complete up until 2012, it appears that the importance of laser scanning technologies in geosciences is still rapidly growing, almost exponentially. We consider that an exponential function is a sensible way to model this growth as the total amount of potential LiDAR users or devices is not infinite and probably another technology
will take over; however the initial stage of growth of a logistic function is approximately exponential, which reasonably fits the observed trend in the Georef database.

![Georef database](image)

**Figure 2.** Number of publications in the Georef database using the term “LiDAR” or “laser scan*” in keywords between 1990 and 2015. These terms were consulted on the Georef database at different dates (see different symbols for 2009, 2014 and 2016). Dashed line: forecast of the number references by [17]. Continuous line: present evolution up to 2012 based on the available catalogue at January 2016 (logistic function fit).

Concerning SFM, one can observe, in Figure 2, the emergence of this technique in the geosciences literature starting from 2011. Based on the January 2016 version of Georef database, the number of records for 2013 and 2014 are strongly underestimated (as for the term “LiDAR”) and will be corrected in the following years. Although SFM has been rapidly adapted by geoscientists, Figure 2 shows that it is at its early stages of development; our opinion is that this technique will become much more widely used in the next five years and that the number of publications using this technique will experience an exponential growth soon. Our approach to forecast has a clear drawback: it is clearly biased by our current knowledge on the previously published articles. Unfortunately, black swan events, i.e., extreme and often unexpected phenomena having the ability to profoundly disrupt the system functioning [50], may also take place during the next 5–10 years. For instance, black swans in the form of new technological developments, are expected to happen during the next decade or so, which may completely dislocate our forecasting attempt. Unfortunately, “prediction is very difficult, especially about the future” as might have said Niels Bohr unless Nostradamus did it [51].

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Abbreviations
The following abbreviations are used in this manuscript:

UAV Unmanned Aerial Vehicles
SfM Structure-from-Motion
TLS Terrestrial Laser Scanner
PIV Particle Image Velocity
HDR High Dynamic Range Images
COPD Cumulative Offset Probability Density
LoD Level of Detection

References
5. Tarolli, P.; Sofia, G.; Dalla Fontana, G. Geomorphic features extraction from high-resolution topography: Landslide crowns and bank erosion. *Nat. Hazards* 2012, 61, 65–83. [CrossRef]
Remote Sens. 2016, 8, 130


37. Tonini, M.; Abellan, A. Rockfall detection from terrestrial LiDAR point clouds: A clustering approach using R. JOSIS 2015, 8, 95–110. [CrossRef]
42. Moulin, A.; Benedetti, L.; Gosar, A.; Rupnik, P.J.; Rizza, M.; Bourlès, D.; Ritz, J.F. Determining the present-day kinematics of the Idrija fault (Slovenia) from airborne LiDAR topography. Tectonophysics 2014, 628, 188–205. [CrossRef]
43. James, M.R.; Robson, S. Straightforward reconstruction of 3D surfaces and topography with a camera: Accuracy and geoscience application. J. Geophys. Res. Earth 2012, 117. [CrossRef]

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