

Interactive comment on “REVIEW ARTICLE: THE USE OF REMOTELY PILOTED AIRCRAFT SYSTEMS (RPAS) FOR NATURAL HAZARDS MONITORING AND MANAGEMENT”

by Daniele Giordan et al.

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Dear Editor,

in the following you can find our answers to referees and the modified manuscript. We also added a number to our answers and you can find the relative modifications in the manuscript.

ANSWER TO REVIEWER 1 and 2 AND TEXT WITH MODIFICATIONS

RC - Referee comments

AC- Author comments – answers have been numbered and in the text and (when possible) labels identified answers to referee comments.

RC - Overall This is a review paper relating to the use of RPAS for natural hazard monitoring and management. It particularly focusing on the use of Mini and Micro RPAS for five kinds of disaster, such as landslides, floods, earthquakes, wildfires and volcano activities. However, the topic and discussed disaster types are similar to the following paper just published last year. Thus, I suggest to major revise this manuscript.

AC1 - We would like to thank the Reviewer for his suggestions. We well know that the topic has also been analyzed by other authors. However, with our review, we tried to make the literature review more complete and updated. We provided more than 150 references considering the most important natural hazards, including some recent articles published in the last year and following the available Annual Disaster Statistical Review. Along these lines, we believe that the natural hazards scientific community will be benefited by such long and updated list of articles and research advances. In detail, the manuscript is focused on the revision of available bibliography for the use of RPAS for: landslides, earthquakes, volcanic activity and wildfires. These four categories are the most dangerous and the manuscript propose a revision of case studies and proposed methodology to fix a possible approach for the use of RPAS in these critical conditions. The description of case studies and possible approaches is important to fix a common methodology that can be used not only for scientific purposes, but also for the management of real emergencies. Until now, the lack of a well-defined methodology that describes pros and cons to the use of RPAS for the support during natural hazard emergencies is a critical aspect that this paper can tries to solve.

RC - Detail comments are stated below. i. An Christopher Gomez and Heather Purdie, 2016, “UAV- ~ based Photogrammetry and Geocomputing for Hazards and Disaster Risk Monitoring – A Review”, Geoenvironmental Disasters, Vol.3, No.23. 2.

Comments i.

The above mentioned article was not referenced, compared or analyzed. It is strongly suggest to include this paper and conduct comparisons to emphasize their different point of view.

AC2 - Line 56: Done. We would like to thank the Reviewer for this suggestion. We added this important paper in our review, and we also considered the bibliography of the manuscript. In our manuscript, we revised 159 papers (100 more than the paper mentioned above) that can be considered an exhaustive representation of available bibliography on this subject. With respect to Gomez and Purdie (2016), we tried to analyze more in-depth the bibliography and define a possible methodology for the use of RPAS that we obtained merging all the revised papers. In the case of landslides, for example, this approach has been used to define the possible use of RPAS for: i) landside recognition, classification and post-event analysis, ii) landslide monitoring, iii) landslide susceptibility and hazard assessment. For each of this points, we present a description of the use of RPAS based on the available bibliography. To clarify this point, we add the following paragraph that introduces the paper of Gomez and Purdie and points out the different approach of our manuscript from line 152 to line 162:

“Gomez and Purdie (2016) published a detailed analysis of the use of RPAS for hazards and disaster risk monitoring. In our paper, we focused our attention on the most dangerous natural hazards that can be analyzed using RPAS. According to the definitions used by Annual Disaster Statistical Review (Guha-Sapir et al., 2016), the paper considers in particular: i) landslides, ii) floods iii) earthquakes v) volcanic activity vi) wildfires. For each considered category of natural hazard, our paper presents a review of a large list of published papers (151 papers) analyzing proposed methodologies and provided results, and underlining strengths and limitations in the use of RPAS. The aim of this paper is the description of the possible use of RPAS in considered natural hazard, describing a general methodology for the use of these systems in different contexts merging all previously published experiences.”

In this revised version of the paper, we also added the following bibliography:

Derrien, A., Villeneuve, N., Peltier, A. and Beauducel, F.: Retrieving 65 years of volcano summit deformation from multitemporal structure from motion: The case of Piton de la Fournaise (La Réunion Island), *Geophys. Res. Lett.*, 42(17), 6959–6966, doi:10.1002/2015GL064820, 2015.

Dewitte, O., J.C. Jasselette, Y. Cornet, M. Van Den Eeckhaut, A. Collignon, J. Poesen, and A. Demoulin.: Tracking landslide displacements by multitemporal DTMs: A combined aerial stereophotogrammetric and LIDAR approach in western Belgium. *Engineering Geology*, 7, 582–586, 2008.

Diaz, J. A., Pieri, D., Wright, K., Sorensen, P., Kline-Shoder, R., Arkin, C. R., Fladeland, M., Bland, G., Buongiorno, M. F., Ramirez, C., Corrales, E., Alan, A., Alegria, O., Diaz, D. and Linick, J.: Unmanned Aerial Mass Spectrometer Systems for In-Situ Volcanic Plume Analysis, *J. Am. Soc. Mass Spectrom.*, 26(2), 292–304, doi:10.1007/s13361-014-1058-x, 2015.

Eugster, H. and Nebiker, S.: UAV-based augmented monitoring– real-time georeferencing and integration of video imagery with virtual globes. In: *Int. Archives of Photogrammetry, Remote Sensing and Spatial Information Sciences*, Beijing, China, 37(B1), 1229–1235. 2008.

Ezequiel, C.A.F., Cua, M., Libatiquem, N.C., Tangonan, G.L., Alampay, R., Labuguen, R.T., Favila, C.M., Honrado, J.L.E., Canos, V., Devaney, C., Loreto, L.B., Bacusmo, J. and Palma, B.: UAV Aerial Imaging Applications for Post-Disaster Assessment, Environmental Management and Infrastructure Development. 2014 International Conference on Unmanned Aircraft Systems (ICUAS) Orlando, FL, USA proceedings: 274–283, 2014.

Fan, J., Zhang, X., Su, F., Ge, Y., Tarolli, P., Yang, Z., Zeng, C., and Zeng, Z.: Geometrical feature analysis and disaster assessment of the Xinmo landslide based on remote sensing data, *Journal of Mountain Science*, 14, 1677–1688, doi:10.1007/s11629-017-4633-3, 2017.

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- Gomez, C.: Digital photogrammetry and GIS-based analysis of the bio-geomorphological evolution of Sakurajima Volcano, diachronic analysis from 1947 to 2006, *J. Volcanol. Geotherm. Res.*, 280, 1–13, doi:10.1016/j.jvolgeores.2014.04.015, 2014.
- Gomez, C. and Kato, A.: Multi-scale voxel-based algorithm for UAV-derived point-clouds of complex surfaces. *IEEE International ICARES – Aerospace Electronics and Remote Sensing Technology*: 205–209. 2014.
- Gomez, C. and Purdie, H.: UAV- based Photogrammetry and Geocomputing for Hazards and Disaster Risk Monitoring – A Review. *Geoenvironmental Disasters* 3(23), 1-11, 2016.
- Gomez, C., Hayakawa, Y. and Obanawa, H.: A study of Japanese landscapes using structure from motion derived DSMs and DEMs based on historical aerial photographs: New opportunities for vegetation monitoring and diachronic geomorphology, *Geomorphology*, 242, 11–20, doi:10.1016/j.geomorph.2015.02.021, 2015.
- Guha-Sapir, D., Hoyois, P., and Below, R.: Annual Disaster Statistical Review 2015 The numbers and trends. Centre for Research on the Epidemiology of Disasters, Ciaco Imprimerie, Louvain-la-Neuve (Belgium), pp. 50, 2016.
- Hervouet, A., Dunford, R., Piégay, H., Belletti, B. and Trémélo, M.-L.: Analysis of Post-flood Recruitment Patterns in Braided-Channel Rivers at Multiple Scales Based on an Image Series Collected by Unmanned Aerial Vehicles, Ultra-light Aerial Vehicles, and Satellites, *GIScience Remote Sens.*, 48(1), 50–73, doi:10.2747/1548-1603.48.1.50, 2011.
- Javernick, L., Brasington, J. and Caruso, B.: Modeling the topography of shallow braided rivers using Structure-from-Motion photogrammetry, *Geomorphology*, 213, 166–182, doi:10.1016/j.geomorph.2014.01.006, 2014.
- Lindner, G., Schraml, K., Mansberger, R. and Hubl J.: UAV monitoring and documentation of a large landslide. *Appl Geomat*, 8(1), 1-11, 2016.
- Liu, C.-C., Chen, P.-L., Tomoya, M., Chen, C.-Y.: Rapidly responding to landslides and debris flow events using a low-cost unmanned aerial vehicle. *J. Rem. Sens.* 9(1), 1-11, doi:10.1117/1.JRS.9.096016, 2015.
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- Nakamura, F., Shimatani, Y., Nishihiro, J., Ohtsuki, K., Itsukushima, R. and Yamada, H.: Report on flood disaster in Kinu River, occurred in September, 2015 (in Japanese with English abstract), *Ecol. Civ. Eng.*, 19(2), 259–267, doi:10.3825/ece.19.259, 2017.
- Nedjati, A., Vizvari, B., Izbirak, G.: Post-earthquake response by small UAV helicopters, *Nat. Hazards* 80, 1669–1688, 2016. Doi: <http://dx.doi.org/10.1007/s11069-015-2046-6>
- Obanawa, H., Y. Hayakawa, and C. Gomez.: 3D Modelling of inaccessible Areas using UAV-based Aerial Photography and Structure from Motion. *Transactions of the Japanese Geomorphological Union*, 35, 283–294. 2014.

Pham, T.-T.-H., P. Apparicio, C. Gomez, C. Weber, and D. Mathon.: Towards a rapid automatic detection of building damage using remote sensing for disaster management. The Haiti earthquake. Dis. Prev. Manage, 23, 53–66, 2014. doi: 10.1108/DPM-12-2012-0148

Pyo, J., Cho, H., Joe, H., Ura, T. and Yu, S.: Development of hovering type AUV “ Cyclops ” and its performance evaluation using image mosaicing, Ocean Eng., 109, 517–530, doi:10.1016/j.oceaneng.2015.09.023, 2015.

Smith, M. W., Carrivick, J. L., Hooke, J. and Kirkby, M. J.: Reconstructing flash flood magnitudes using “Structure-from-Motion”: A rapid assessment tool, J. Hydrol., 519, 1914–1927, doi:10.1016/j.jhydrol.2014.09.078, 2014.

Thamm, H.P. and Judex, M.: The “Low cost drone” – An interesting tool for process monitoring in a high spatial and temporal resolution. The International Archives of Photogrammetry, Remote Sensing and Spatial Information Sciences, Enschede, The Netherlands, Vol. XXXVI part 7. 2006

Tamminga, A. D., Eaton, B. C. and Hugenholtz, C. H.: UAS-based remote sensing of fluvial change following an extreme flood event, Earth Surf. Process. Landforms, 40(11), 1464–1476, doi:10.1002/esp.3728, 2015.

Witek, M., Jeziorska, J. and Niedzielski, T.: An experimental approach to verifying prognoses of floods using an unmanned aerial vehicle, Meteorol. Hydrol. Water Manag., 2(1), 3–11 [online] Available from: <http://www.mhwm.pl/An-experimantal-approach-to-verifying-prognoses-of-floods-using-unmanned-aerial-vehicle,0,8.html>, 2014.

Woodget, A. S., Carbonneau, P. E., Visser, F. and Maddock, I. P.: Quantifying submerged fluvial topography using hyperspatial resolution UAS imagery and structure from motion photogrammetry, Earth Surf. Process. Landforms, 40, 47–64, doi:10.1002/esp.3613, 2015.

Xie, Z., J. Yang, C. Peng, Y. Wu, X. Jiang, R. Li, Y. Zheng, Y. Gao, S. Liu, and B. Tian.: Development of an UAS for post-earthquake disaster surveying and its application in Ms7.0 Lushan Earthquake, Sichuan, China. Comput. Geosc. 68, 22–30, 2014.

RC - ii. The used acronyms are not consistent, RPAS, UAV, UAS, UVS were adopted at different places of the paper. If their definitions have major difference, the authors should define them clearly. If not, using one acronym for the whole paper may be considered.

AC3 - We revised the text, and we corrected these discrepancies.

RC - iii. Line 28-31, numbers within () should include unit, such as 380, 22765, etc. iv. Line 48, what is RLS and what are RTK/PPK at Line 95? The first time an acronym appear, its whole name should be explained. On the contrary, the explanation of GCP appear twice in the paper.

AC4 - We improve the text according to reviewer’s suggestions. In particular:

From line 29 – we updated all the values of the ADSR with unit. We also added the average number of events per year (we add events in the text)

22,765 are fatalities (we add in the text)

RLS should read SLR (single lens reflex) camera (we correct the text)

RTK is Real Time Kinematic whereas PPK is Post Processing Kinematic (we add in the text)

RC - v. Table 1 specify the classification of Mini/Micro UAV. A reference should be referred.

AC5 - UVS International definition, added

RC - vi. Line 476, "small UAV" is used. What is its definition?

AC6- fixed

RC - vii. Meanwhile, I doubt the definition in Table 1 is correct, as the Max. Flight altitude for Micro UAV is FIXED at 250m and its endurance time is also FIXED at 1h.

AC7 - Flight altitude depends on countries whereas endurance depends on the payload. Reported numbers are just indicative.

RC - viii. In this paper, the authors focus on the use of Mini and Micro RPAS only. However, these two kinds of RPAS are not suitable for volcano activities study, because its maximum flight altitude is generally lower than a volcano. For example at Line 422, a fixed-wing UAV can fly over Mt. Etna up to 4000m. This fixed-wing is not belong to the Mini or Micro RPAS. Right? There are other similar case studies that didn't use Mini or Micro RPAS as well.

AC8 - We thank the reviewer for this important issue. We added in the text that, for volcanoes, we also considered larger RPAS:

Line 431: "As mentioned before, this paper considers in particular small RPAS. In the study of volcanoes, larger aircrafts with a payload of kilograms are also utilized to mount other types of sensors to monitor various aspects of their dynamic activities. For this reason, in this chapter, we also consider larger RPAS solutions."

RC - ix. Line 212, RPAs or RPAS?.

AC9 - RPAS

RC - x. Line 415, two references for Gomez are not found in the list of reference.

AC10 - We added missing references, and we made a cross-check of all references mentioned in the manuscript and published in the reference list

ANSWER TO REVIEWER 2

RC - Overall This is a review paper relating to the use of small RPAS for natural hazards monitoring and management for five kinds of disasters, such as landslides, floods, earthquakes, wildfires and volcanos. The paper recites many international papers and summarizes their content and results briefly. The focus is on the use of small RPAS (<30 kg MTOW) in combination with optical sensor systems (mainly), laser scanners and gas detection systems. The introduction explains the two classes of RPAS and the common workflow

of using an RPAS and post post-processing the aerial single images or video streams (nadir and oblique view) by using common Structure from Motion Software Tools (like Pix4, AgiSoft, Capturing Reality, DroneDeploy, etc.) to generate data products like orthophotos and point clouds. The advantages of using RPAS for natural hazards assessment are well described related to the use of aerial camera systems (for RGB, Multi-/Hyperspectral and TIR range). Possible accuracies of these data products are described too in dependence of using GCPs, a low cost AHRS and/or high end GNSS/INS system in combination with the optical sensor system. This paper is a good introduction to the usage of RPAS for natural hazards monitoring and even latest results are listed - i.e. using deep learning algorithms / CNN for detecting destroyed facades to provide relevant information on-site and in near real time for first responders (section 2.3).

AC11 – we would like to thank the Reviewer for the good description of the paper that shows several important issues considered.

RC - Sadly, there are no recommendations for best practices or open source tools and no comparison or rating of the described workflows of each section (landslides, floods, earthquakes, wildfires, volcanos). Especially for using SfM-Software many publications are available which analyses image processing time, achievable accuracies of resulting data products by using / not using GCPs, alternating flight strips and/or cross strips and AHRS or GNSS/INS solutions and the effects of using a metric or non-metric camera system - i.e. DJI Phantom 4 Pro (metric) and DJI Mavic (sadly not metric).

AC12 – We thank the reviewer for this suggestion. In this paper, we decided to focus our attention on natural hazards and possible use of RPAS. The analysis of available bibliography shows that the possible solutions are so different and dependent from the final goal of the mission and the end users requirements that is quite impossible to propose a generic workflow for each natural hazard. For this reason, we decided to propose a generic workflow in chapter one (figure 2) and then propose a large analysis of available bibliography for each analyzed natural hazard. We also decided to do not compare software or RPAS performances because it was not the aim of the paper and we don't consider the comparison of available software a burning research topic as similar papers have been already published in the past (see Remondino et al., 2014 in Photogrammetric Record). We followed the requests of reviewer 2 and we added several sentences, in particular:

From line 108 to line 113: “The use of GCP and different GNSS solutions is a fundamental point. Gerke and Przybilla (2016) presented the effect of RTK-GNSS and cross flight patterns, and Nocerino et al., (2013) presented an evaluation about RPAS processing results quality considering: i) the use of GCPs, ii) different photogrammetric procedures, iii) different network configurations. If a quick mapping is needed, the information delivered by the navigation system can be directly used to stitch the images and produce a rough image mosaicking (Chang-chun et al., 2011).”

RC - Comments Line No. 27: You cite the Annual Disaster Statistical Review of 2015. The Citation ADSR, 2015 is missing in the reference section and I suggest to update the statistic numbers by using the latest report of 2016.

AC13 – At the moment of submission, ADSR 2016 was not available. Now we updated with this publication.

ADSR 2015 was already cited in bibliography as now ADSR 2016, with the suggested citation: Guha-Sapir, D., Hoyois, P., Wallemacq P. and Below, R.: Annual Disaster Statistical Review 2016 The numbers and trends.

Centre for Research on the Epidemiology of Disasters, Ciaco Imprimerie, Louvain-la-Neuve (Belgium), pp. 91, 2017

RC - Line No. 37: You address a crucial point here. Time matters, especially during the disaster assessment or disaster monitoring phase. With a RPAS you are easily able to monitor on-site in real time. Why is there no section in your paper where you discuss reliable or suitable RPAS solutions compared to common satellite based solutions / services. There is also another issue to be mentioned. Capturing high res images or videos can be done on time but the main bottleneck is the time which is necessary to post-process that huge amount of images (i.e. with SfM Tools) to generate maps, mosaics, orthophotos, point clouds etc. Several case studies have been published by <http://drones.fsd.ch/en/> which should be considered to take into account.

AC14 – we thank the reviewer for this suggestion and we added these two paragraphs:

From line 62 to line 70: Another important added value of RPAS is their adaptability that allows their use in various typologies of missions, and in particular for monitoring operations in remote and dangerous areas (Obanawa et al., 2014). The possibility to carry out flight operations at lower costs compared to ones required by traditional aircraft is also a fundamental advantage. Limited operating costs make these systems also convenient for multi-temporal applications where it is often necessary to acquire information on an active process (like a landslide) over the time. Beside their higher resolution and the possibility to extract reliable 3D information, UAV images are not conditioned by cloud cover as satellite imagery. A comparison between the use of satellite images, traditional aircraft and RPAS has been presented and discussed by Fiorucci et al. (2018) for landslides applications and by Giordan et al., (2018) for the identification of flooded areas. These contributions demonstrated the goodness of RPAS for on demand acquisitions of high resolution images over limited areas.

from line 546 to line 549: “In particular during emergencies, the time required for RPAS dataset processing is an important element that should be carefully considered. Giordan et al. (2015a) presented a case study related to a landslide emergency. In this paper, authors considered not only possible results but also the time that is required for them.”

RC -Line No. 45: "contest" or "context" of remote sensing research?

AC15 – context

RC - Line No. 48: SLR instead of RLS. I suggest to replace by "integrated camera systems" as well to address all kind of optical solutions for RPAS (i.e. bridge cameras, industrial grade cameras, video cameras, etc.).

AC16 – we modified the sentence from line 49 to 51: “In particular, the development of photogrammetry and technologies associated (i.e. integrated camera systems like compact cameras, industrial grade cameras, video cameras, single-lens reflex (SLR) digital cameras and GNSS/INS systems) allow to use of RPAS platforms in various applications as alternative to the traditional remote sensing method for topographic mapping or detailed 3D recording of ground information and a valid complementary solution to terrestrial acquisitions too (Nex and Remondino, 2014) (Fig.1).”

RC - Section from Line No. 52 to 62: I recommend to add the advantage of "micro RPAS are easy to transport into the disaster area". Foldable Systems (like DJI Mavic) fits easily into a day pack and can be transported safely as hand luggage. Weight matters especially for first responder teams like UNDAC or similar.

AC17 – line 56 to line 60: RPAS systems present some advantages in comparison to traditional platforms and, in particular, they could be competitive thanks to their versatility in the flight execution (Gomez and Purdie, 2016). Mini/micro RPAS are the most diffused for civil purposes, and they can fly at low altitudes according to limitations defined by national aviation security agencies and be easy transported into the disaster area. Foldable systems fits easily into a daypack and can be transported safely as hand luggage. This advantage is particularly important for first responder teams like UNDAC or similar.

RC - Section from line no. 83 to 104: I recommend to add some references to papers which analyses possible accuracies by using / not using GCPs and SFM Tools (i.e. Pix4D, Agisoft) or common photogrammetric workflows (i.e. Inpho Match AT). I suggest as well to add some references here to fast mosaicking methods - i.e. PhaseOne and IGI showed promising results with the commercial IGI Mapper System and the German Aerospace Center developed specialized solutions for realtime traffic management (VABENE) and realtime mapping applications (MACS) on manned and unmanned aircrafts. Intro section in general: You name laser scanning and gas detection and also reference on that in section 2.1.1, 2.1.2, 2.1.3, 2.4.2 and 2.5 but a workflow description is missing. I recommend to add this workflow description or to specify the argumentation of using optical sensor systems.

AC18 – as we mentioned before, the principal aim of this manuscript is a review of available bibliography. We decided to avoid the publication of performance comparison between RPAS and/or software because we think that the focus is different. We mentioned papers like Remondino et al., (2014) and Nocerino et al., (2015) that considered this topic to complete our review. We thank for the suggestion about the rapid mapping and we added the following sentence (lines 124-127): "In particular during emergencies, the time required for the image dataset processing can be a critical point. For this reason, the development of fast mosaicking methods as MACS, for a real time mapping applications, or VABENE++, developed by German Aerospace Center for real time traffic management (Detzer et al., 2015)."

RC - Line No. 128: Reference of (ADSR 2015) is missing. Update to ADSR 2016 is recommended.

AC19 – ADSR 2015 was already cited in bibliography as now ADSR 2016, with the suggested citation: Guha-Sapir, D., Hoyois, P., Wallemacq P. and Below, R.: Annual Disaster Statistical Review 2016 The numbers and trends. Centre for Research on the Epidemiology of Disasters, Ciaco Imprimerie, Louvain-la-Neuve (Belgium), pp. 91, 2017

RC - Section 2.1: I recommend to add the main parameters which influence the accuracy of derived DEM and orthophotos (i.e. real GSD, knowledge about interior and exterior orientation parameters, overlap of images, flight strip configuration and used SfM-Software)

AC20 – we added (line 120-123): In real applications, many parameters can influenced the final resolution of DSM/DTM and orthophoto like: real GSD (Nocerino et al., 2013) interior and exterior orientation parameters (Kraft et al., 2016), overlap of images, flight strip configuration and used SfM-MVS software (Nex et al., 2015).

Line No. 281: First use of SfM-MVS - please explain.

AC21 – Structure from Motion-Multi View Stereo (SfM-MVS), we improved the text

REVIEW ARTICLE: THE USE OF REMOTELY PILOTED AIRCRAFT SYSTEMS (RPAS) FOR NATURAL HAZARDS MONITORING AND MANAGEMENT

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ABSTRACT

The number of scientific studies that consider possible applications of Remotely Piloted Aircraft Systems (RPAS) for the management of natural hazards effects and the identification of occurred damages are strongly increased in last decade. Nowadays, in the scientific community, the use of these systems is not a novelty, but a deeper analysis of literature shows a lack of codified complex methodologies that can be used not only for scientific experiments but also for normal codified emergency operations. RPAS can acquire on-demand ultra-high resolution images that can be used for the identification of active processes like landslides or volcanic activities but also for the definition of effects of earthquakes, wildfires and floods. In this paper, we present a review of published literature that describes experimental methodologies developed for the study and monitoring of natural hazards.

1. INTRODUCTION

In last three decades, the number of natural disasters showed a positive trend with an increase in the number of affected populations. Disasters not only affected the poor and characteristically more vulnerable countries but also those thought to be better protected. Annual Disaster Statistical Review describes recent impacts of natural disasters over population and reports 376342 natural triggered disasters in 2015 (ADSR, 2015), 2016 (Guha-Sapir et al., 2017). This is less than the annual average annual-disaster frequency observed from 20052006 to 2014 (3802015 (376.4 events)), however natural disasters is still responsible for a high number of casualties (22,7658,733 death). In 2015the period 2006-2015, the average number of casualties annaly caused by natural disasters is 69,827. In 2016, hydrological disasters (175177) had the largest share in natural disaster occurrence (46.551.8%), followed by meteorological disasters (127; 33.896; 28.1%), climatological disasters (45; 1238; 11.1%) and geophysical disasters (29; 7.7%) (ADSR, 201531; 9.1%) (Guha-Sapir et al., 2017). To face these disasters, one of the most important solutions is the use of systems able to provide an adequate level of information for correctly understanding these events and their evolution. In this context, survey and monitoring of natural hazards gained in importance. In particular, during the emergency phase it is very important to evaluate and control the phenomenon evolution, preferably operating in near real time or real time, and consequently, use this information for a better risk scenario assessment. The available acquired data must be processed rapidly to ensure the emergency services and decision makers promptly.

Commentato [G1]: AC13

Commentato [G2]: AC2 -

40 Recently, the use of remote sensing (satellite and airborne platform) in the field of natural hazards and
41 disasters has become common, also supported by the increase in geospatial technologies and the ability to
42 provide and process up-to-date imagery (Joyce et al., 2009; Tarolli, 2014). Remotely sensed data play an
43 integral role in predicting hazard events such as floods and landslides, subsidence events and other ground
44 instabilities. Because their acquisition mode and capability for repetitive observations, the data acquired at
45 different dates and high spatial resolution can be considered as an effective complementary tool for field
46 techniques to derive information on landscape evolution and activity over wide areas.

47 In the contest of remote sensing research, recent technological developments have increased in the field of
48 Remotely Piloted Aircraft Systems (RPAS) becoming more common and widespread in civil and commercial
49 context (BoccardoBendea et al., 2008). In particular, the development of photogrammetry and technologies
50 associated (i.e. RLSintegrated camera systems like compact cameras, industrial grade cameras, video
51 cameras, single-lens reflex (SLR), digital cameras and GNSS/INS systems) allow to use of RPAS platforms in
52 various applications as alternative to the traditional remote sensing method for topographic mapping or
53 detailed 3D recording of ground information and a valid complementary solution to terrestrial acquisitions
54 too (Nex and Remondino, 2014) (Fig.1).

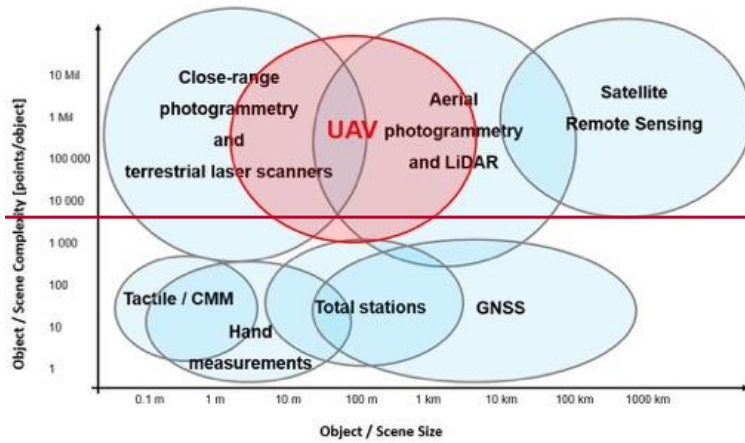
55 RPAS systems present some advantages in comparison to traditional platforms and, in particular, they could
56 be competitive thanks to their versatility in the flight execution: (Gomez and Purdie, 2016). Mini/micro RPAS
57 are the most diffused for civil purposes, and they can fly at low altitudes according to limitations defined by
58 national aviation security agencies: and be easy transported into the disaster area. Foldable Systems fits
59 easily into a daypack and can be transported safely as hand luggage. This advantage is particularly important
60 for first responder teams like UNDAC or similar. Stöcker et al. (2017) published a review of different state
61 regulations that are characterized by several differences regarding requirements, distance from the takeoff
62 and maximum altitude. Another important added value of RPAS is their adaptability that allows their use in
63 various typologies of missions, and in particular for monitoring operations in remote and dangerous areas:
64 (Obanawa et al., 2014). The possibility to carry out flight operations at lower costs compared to ones required
65 by traditional aircraft is also a fundamental advantage. Limited operating costs make these systems also
66 convenient for multi-temporal applications where it is often necessary to acquire information on an active
67 process (like a landslide) over the time. A comparison between the use of satellite images, traditional aircraft
68 and RPAS has been presented and discussed by Fiorucci et al. (2018) for landslides applications and by
69 Giordan et al., (2017) for the identification of flooded areas. These comparisons show that RPAS are a good
70 solution for the on demand acquisition of high resolution images over limited areas.

Commentato [G3]: AC16

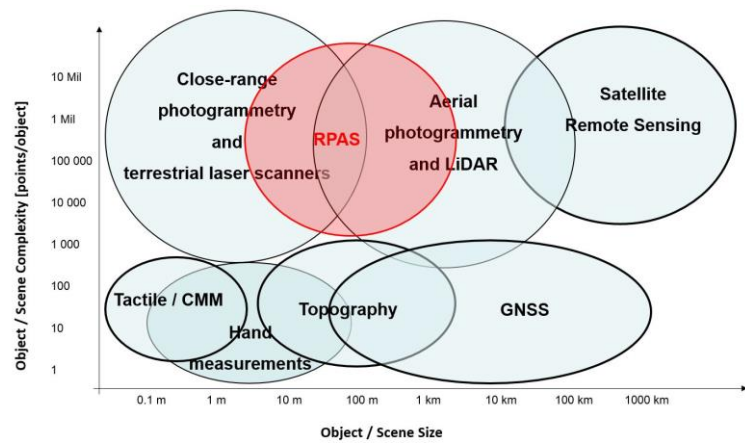
Commentato [G4]: AC2

Commentato [G5]: AC17

Commentato [G6]: AC14



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73 Figure 1. Available geomatics techniques, sensors, and platforms for topographic mapping or detailed 3D
 74 recording of ground information, according to the scene dimensions and complexity (modified from Nex and
 75 Remondino, 2014).

76 RPASs are used in several fields as agriculture, forestry, archaeology and architecture, traffic monitoring,
 77 environment and emergency management. In particular, in the field of emergency assistance and
 78 management, RPAS platforms are used to reliably and fast collect data of inaccessible areas. Collected data
 79 can be mostly images but also gas concentrations or radioactivity levels as demonstrated by the tragic event
 80 in Fukushima ([Sanda et al., Sanada and Torii, 2015](#); [Martin et al., 2016](#)). Focusing on image collection, they
 81 can be used for early impact assessment, to inspect collapsed buildings and to evaluate structural damages
 82 [on common infrastructures](#) ([Chou et al. 2010](#); [Molina et al. 2012](#); [Murphy et al., 2008](#); [Pratt et al., 2009](#)) [or](#)
 83 [cultural heritage sites](#) ([Pollefeys et al., 2001](#); [Manfredini et al., 2012](#); [Koutsoudisa et al., 2014](#); [Lazzari et al.,](#)
 84 [2017](#)). Environmental and geological monitoring can profit from fast multi-temporal acquisitions delivering
 85 high-resolution images ([Thamm and Judex 2006](#); [Niethammer et al. 2010](#)). RPAS can be considered a good
 86 solution also for mapping and monitoring different active processes at the earth surface ([Fonstad et al., 2013](#);

87 [Piras et al., 2017](#)) such as: glaciers (Immerzel et al., 2014, Ryan et al., 2015), Antarctic moss beds (Lucieer et
88 al., 2014b), coastal areas (Delacourt et al., 2009; Klemas, 2015), [Interseismic deformations \(Deffontaines et](#)
89 [al., 2017\)](#), river morphodynamic (~~Jaud et al.,~~ [Gomez and Purdie, 2016](#)); [Jaud et al., 2016](#); [Aicardi et al., 2017](#);
90 [Bolognesi et al., 2016](#)), [debris flows \(Wen et al., 2011\)](#), and river channel vegetation (Dunford et al., 2009).

91 The incredible diffusion of RPAS has pushed many companies to develop dedicated sensors for these
92 platforms. Besides the conventional RGB cameras other camera sensors are nowadays available on the
93 market. Multi- and hyper-spectral cameras, as well as thermal sensors, have been miniaturized and
94 customized to be hosted on many platforms.

95 The general workflow of a UAV acquisition is presented in Figure 2 below. The resolution of the images, the
96 extension of the area as well as the goal of the flight are the main constraints that affect the selection of the
97 platform and the typology of the sensor. Large areas can be flown using fixed wing (or hybrid) solutions able
98 to acquire nadir images in a fast and efficient way. Small areas or complex objects (like steep slopes or
99 buildings) should be acquired using rotor RPAS as they are usually slower but they allow the acquisition of
100 oblique views. If the information different from the visible band is needed, the RPAS can host one or more
101 sensors acquiring in different bands. The flight mission can be planned using dedicated software: they range
102 from simple apps installed on smartphones in the low-cost solutions, to laptops connected to directional
103 antennas and remote controls for the most sophisticated platforms. According to the typology of the
104 platform, different GNSS and IMU can be installed. Low-cost solutions are usually able to give positions with
105 few meters accuracy and need GCP (Ground Control Points) to geo-reference the images. On the other hand,
106 most expensive solutions install double frequency GNSS receivers with the possibility to get accurate geo-
107 referencing thanks to ~~RTK or PPK corrections-~~ [Real Time Kinematic \(RTK\) or Post Processing Kinematic \(PPK\)](#)
108 [corrections. The use of GCP and different GNSS solutions is a fundamental point. Gercke and Przybilla \(2016\)](#)
109 [presented the effect of RTK-GNSS and cross flight patterns, and Nocerino et al., \(2013\) presented an](#)
110 [evaluation about RPAS processing results quality considering: i\) the use of GCPs, ii\) different](#)
111 [photogrammetric procedures, iii\) different network configurations.](#) If a quick mapping is needed, the
112 information delivered by the navigation system can be directly used to stitch the images and produce a rough
113 image mosaicking- ([Chang-chun et al., 2011](#)). In the alternative, the typical photogrammetric process is
114 followed: (i) image orientation, (ii) DSM generation and (iii) orthophoto generation. The position (geo-
115 referencing) and the attitude (rotation towards the coordinates system) of each acquisition is obtained by
116 estimating the image orientation. In the dense point cloud generation, 3D point clouds are generated from a
117 set of images, while the orthophoto is generated in the last step combining the oriented images projected
118 on the generated point cloud, leading to orthorectified images- ([Turner et al., 2012](#)). Point clouds can be very
119 often converted in Digital Surface Models (DSM), and Digital Terrain Models (DTM) can be extracted
120 removing the off ground regions (mainly buildings and trees). [In real applications, many parameters can](#)
121 [influenced the final resolution of DSM/DTM and ortophoto like: real GSD \(Nocerino et al., 2013\) interior and](#)
122 [exterior orientation parameters \(Kraft et al., 2016\), overlap of images, flight strip configuration and used](#)
123 [SfM-Software \(Nex et al., 2015\).](#)

Commentato [G7]: AC12

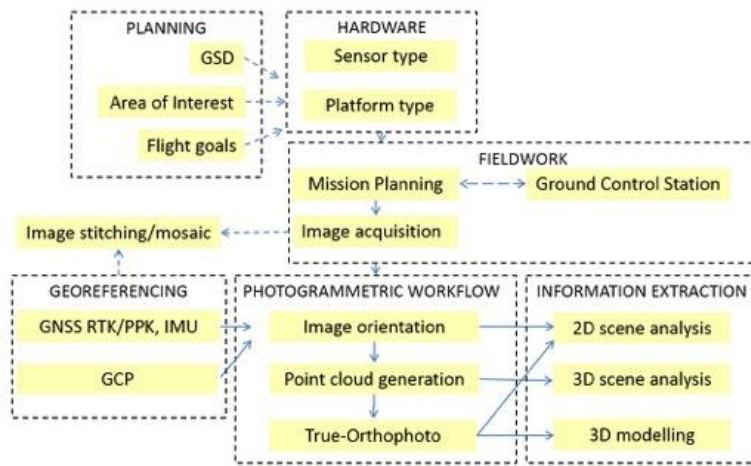
124 [In particular during emergencies, the time required for the image dataset processing can be a critical point.](#)
125 [For this reason, the development of fast mosaicking methods as MACS, for a real time mapping applications](#)
126 [\(Lehmann et al., 2011\), or VABENE++, developed by German Aerospace Center for real time traffic](#)
127 [management \(Detzer et al., 2015\).](#)

Commentato [G8]: AC21

Commentato [G9]: AC18

128 The outputs from the last two steps (point clouds and true-orthophotos) as well as the original images are
129 very often used as input in the scene understanding process: classification of the scene or extraction of

130 features (i.e. objects) of interest using machine learning techniques are the most common applications. 3D
 131 models can also be generated using the point cloud and the oriented images to texturize the model.



132
 133 Figure 2. Acquisition and processing of RPAS images: general workflow.

134 In this paper, the authors present an analysis and evaluation concerning the use of RPAS as alternative
 135 monitoring technique to the traditional methods, relating to the natural hazard scenarios. The main goal is
 136 to define and test the feasibility of a set of methodologies that can be used in the monitoring and mapping
 137 activities. The study is focused in particular on the use of mini and micro RPAS systems (Table 1). The
 138 following table listed the technical specifications of these two RPAS categories, again based on the current
 139 classification by UVS (Unmanned Vehicle Systems) International. Most of the mini or micro RPAS systems
 140 available integrate a flight control system, which autonomously stabilizes these platforms and enables the
 141 remotely controlled navigation. Additionally, they can integrate an autopilot, which allows an autonomous
 142 flight based on predefined waypoints. For the monitoring and mapping applications, mini- or micro RPAS
 143 systems are very useful as cost-efficient platforms for capturing real-time close-range imagery. These
 144 platforms can reach the area of investigation and take several photos and videos from several points and
 145 different angles of view. (Gomez and Kato, 2014). For mapping applications, it is also possible to use this flight
 146 control data to geo-register the captured payload sensor data like still images or video streams (Eugster and
 147 Nebiker, 2008).

148 Table 1. Classification of mini and micro UAV systems, according to UVS International.

Category	Max. Take Of Weight	Max. Flight Altitude	Endurance	Data Link Range
Mini	<30kg	150-300m	<2h	<10km
Micro	<5Kg	250m	1h	<10km

149
 150 **2. USE OF RPAS FOR NATURAL HAZARDS DETECTION AND MONITORING**
 151

152 Gomez and Purdie (2016) published a detailed analysis of the use of RPAS for hazards and disaster risk
153 monitoring. In our paper, we focused our attention on the most dangerous natural hazards that can be
154 analyzed using RPAS. According to the definitions used by Annual Disaster Statistical Review (ADSR,
155 2015 Guha-Sapir et al., 2017), the paper considers ~~in particular phenomena that can be analyzed using RPAS~~
156 ~~and~~ in particular: i) landslides, ii) floods iii) earthquakes v) volcanic activity vi) wildfires. For each considered
157 category of natural hazard, the paper presents ~~an analysis~~ a review of a large list of published papers (159
158 papers), analyzing proposed methodologies and ~~provide~~ provided results, and underlining strengths and
159 limitations in the use of RPAS. The aims of this paper is the description of possible use of RPAS in considered
160 natural hazards, describing a general methodology for the use of these systems in different contexts merging
161 all previous published experiences.

Commentato [G10]: AC2

162 2.1 Landslides

163 Landslides are one of the major natural hazards that produce each year enormous property damage
164 regarding both direct and indirect costs. Landslides are rock, earth or debris flows on slopes due to gravity.
165 The event can be triggered by a variety of external elements, such as intense rainfall, water level change,
166 storm waves or rapid stream erosion that cause a rapid increase in shear stress or decrease in shear strength
167 of slope-forming materials. Moreover, the pressures of increasing population and urbanization, human
168 activities such as deforestation or excavation of slopes for road cuts and building sites, etc., have become
169 important triggers for landslide occurrence. Because the factors affecting landslides can be geophysical or
170 human-made, they can occur in developed and undeveloped areas.

172 In the field of natural hazards, the use of RPAS for landslides study and monitoring represents one of the
173 most common applications. The number of papers that present case studies or possible methodologies
174 dedicated to this topic has strongly increased in last few years and now the available bibliography offers a
175 good representation of possible approaches and technical solutions.

176 When a landslide occurs, the first information to be provided is the extent of the area affected by the event
177 (figure 3). The landslide impact extent is usually done based on detailed optical images acquired after the
178 event. From these acquisitions, it is possible to derive Digital Elevation Models (DEMs) and orthophotos that
179 allow detecting main changes in geomorphological figures- (Fan et al., 2017). In this scenario, the use of the
180 mini-micro RPAS is practical for small areas and optimal for landslides that often cover an area that range
181 from less than one square kilometres up to few square kilometres. Ultra-high resolution images acquired by
182 RPAS can support the definition not only of the identification of studied landslide limit, but also the
183 identification and mapping of main geomorphological features (Fiorucci et al., 20172018). Furthermore, a
184 sequence of RPAS acquisitions over the time can provide useful support for the study of the gravitational
185 process evolution.

186 According to Scaioni et al. (2014), applications of remote sensing for landslides investigations can be divided
187 into three classes: i) landslide recognition, classification and post-event analysis, ii) landslide monitoring, iii)
188 landslide susceptibility and hazard assessment.



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Figure 3. Example of RPAS image of a rockslide occurred on a road. The image was acquired after the rockslide occurred in 2014 in San Germano municipality (Piemonte region, NW Italy). As presented in Giordan et al. (2015a), a multi-rotor of local Civil Protection Agency was used to evaluate occurred damages and residual risk. RPAS images can be very useful to have a representation from a different point of view of the occurred phenomena. Even not already processed using SFM applications, this dataset can be very useful for decision makers to define the strategy for the management of the first phase of emergency.

2.1.1 Landslides recognition

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The identification and mapping of landslides are usually performed after intense meteorological events that can activate or reactivate several gravitational phenomena. The identification and mapping of landslides can be organized in landslides event maps. Landslides event mapping is a well-known activity obtained through field surveys (Santangelo et al., 2010), visual interpretation of aerial or satellite images (Brardinoni et al., 2003; Ardizzone et al., 2013) combined analysis of LiDAR DTM and images (Van Den Eeckhaut et al., 2007; Haneberg et al., 2009; Giordan et al., 2013; Razak et al., 2013; Niculita et al., 2016). The use of RPAS for the identification and mapping of a landslide has been described by several authors (Niethammer et al. 2009; Niethammer et al. 2010; Rau et al., 2011; Carvajal et al., 2012; Travalletti et al., 2012; Torrero et al., 2015; Casagli et al., 2017). Niethammer et al. (2009) and Liu et al. (2015) showed how RPAS could be considered a good solution for the acquisition of ultra-high resolution images with low-cost systems. Fiorucci et al. (2017) compared the results of the landslide limit mapped using different techniques and found that satellite images can be considered a good solution for the identification and map of landslides over large areas. On the contrary, if the target of the study is the definition of landslide's morphological features, the use of more detailed RPAS images seemed to be the better solution. As suggested by Walter et al., (2009) and Huang et al., (2017) one of the most critical elements for a correct georeferencing of acquired images are the use of Ground Control Points (GCPs). The in situ installation and positioning acquisition of GCPs can be an important challenge in particular in dangerous areas as active landslides. Very often, GCPs are not

217 installed in the most active part of the slide but on stable areas. This solution can be safer for the operator,
218 but it can also reduce the accuracy of the final reconstruction.

219 Another parameter that can be considered during the planning of the acquisition phase is the morphology of
220 the studied area. According to with Giordan et al., (2015b), slope materials and gradient can affect the flight
221 planning and the approach used for the acquisition of the RPAS images. Two possible scenarios can be
222 identified: i) steep to vertical areas (>40°); ii) slopes with gentle to moderate slopes (<40°). In the first case,
223 the use of multi-copters with oblique acquisitions is often the best solution. On the contrary, with more
224 gentle slopes, the use of fixed-wing systems can assure the acquisition of wider areas.

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226 **2.1.2 Landslides monitoring**

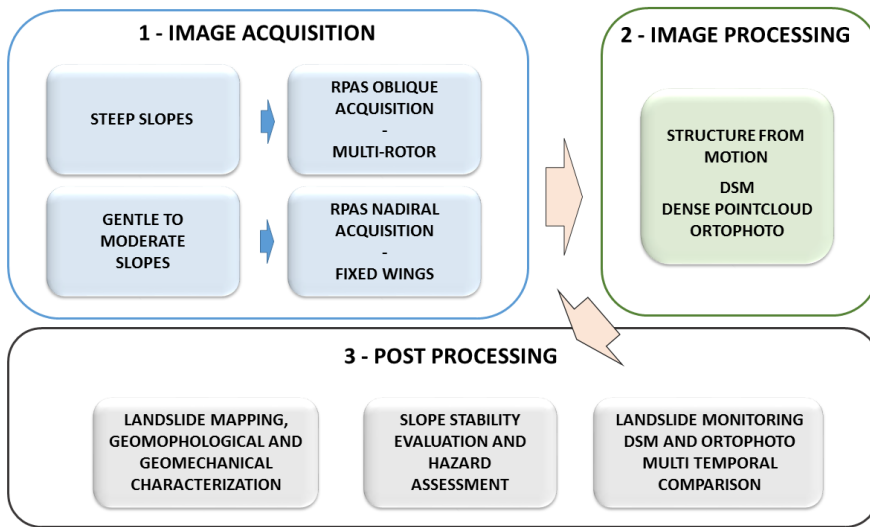
227 The second possible field of application of RPAS is the use of multi-temporal acquisitions for landslides
228 monitoring. This topic has been described by several authors ([Dewitte et al., 2008](#); Turner and Lucieer, 2013;
229 [Travelletti et al., 2012](#); [Lucieer et al. 2014a](#); Turner et al., 2015; Marek et al., 2015; [Lindner et al., 2016](#)). In
230 these works, numerous techniques based on the multi-temporal comparison of RPAS datasets for the
231 definition of the evolution of landslides have been presented and discussed. Niethammer et al. (2010 and
232 2012) described how the position change of geomorphological features (in particular fissures) could be
233 considered for a multi-temporal analysis with the aim of the characterization of the landslide evolution.
234 Travelletti et al. (2012) introduced the possibility of a semi-automatic image correlation to improve this
235 approach. The use of image correlation techniques has been also described by Lucieer et al. (2014a) who
236 demonstrated that COSI-Corr (Co-registration of Optically Sensed Imaged and Correlation - ~~LePrince~~[LePrince](#)
237 et al. 2007, ~~2009~~[2008](#); Ayoub et al., 2009) can be adopted for the definition of the surface movement of the
238 studied landslide. A possible alternative solution is the multi-temporal analysis of the use of DSMs. The
239 comparison of digital surface models can be used for the definition of volumetric changes caused by the
240 evolution of the studied landslide. The acquisition of these digital models can be done with terrestrial laser
241 scanners (Baldo et al., 2009) or airborne LiDAR (Giordan et al., 2013). Westoby et al. (2012) emphasized the
242 advantages of ~~RPAs~~[RPAS](#) concerning terrestrial laser scanner, which can suffer from line-of-sight issues, and
243 airborne LiDAR, which are often cost-prohibitive for individual landslide studies. Turner et al. (2015) stressed
244 the importance of a good co-registration of multi-temporal DSM for good results that could decrease the
245 accuracy of results. The use of benchmarks in areas not affected by morphological changes can be used for a
246 correct calibration of rotational and translation parameters.

247

248 **2.1.3 Landslides susceptibility and hazard assessment**

249 Landslides susceptibility and hazard assessment are often performed at basin scale (Guzzetti et al., 2005)
250 using different remote sensing techniques (Van Westen et al., 2008). The use of RPAS can be considered for
251 single case study applications to help decision makers in the identification of the landslide damages and the
252 definition of residual risk (Giordan et al., 2015a). Saroglou et al., (2017) presented the use of RPAS for the
253 definition of trajectories of rock falls prone areas. Salvini et al. (2017) and Török et al., (2017) described the
254 combined use of TLS and ~~RPAs~~[RPAS](#) for hazard assessment of steep rock walls. All these papers considered
255 the use of RPAS as a valid solution for the acquisition of DSM over sub-vertical areas. Török et al., (2017) and
256 Tannant et al., 2017 also described in their manuscripts how RPAS DSMs can be used for the evaluation of
257 slope stability using numerical modelling. [Fan et al. \(2017\) analyzed the geometrical features and provided](#)

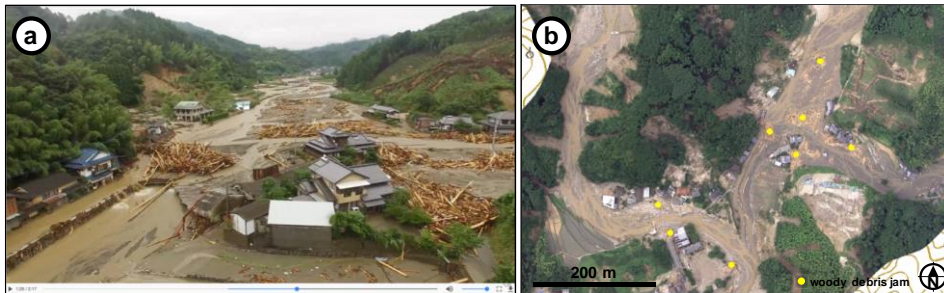
258 the disaster assessment of a landslide occurred on June 24 2017 in the village of Xinmo in Maoxian County,
 259 (Sichuan Province, Southwest China). Aerial images were acquired the day after the event from an unmanned
 260 aerial vehicle (UAV) (fixed-wing UAV, with a weight less than 10 kg, and flight autonomy up to 4 hours), and
 261 a digital elevation model (DEM) was processed, with the purpose to analyzed the main landslide geometrical
 262 features (front, rear edge elevation, accumulation area, horizontal sliding distance)



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 265 Figure 4. Acquisition, processing and post-processing of RPAS images applied to i) landslides recognition, ii)
 266 hazard assessment and iii) slope evolution monitoring.
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268 **2.2 Floods**

269 Disastrous floods in urban, lowland areas often cause fatalities and severe damage to the infrastructure.
 270 Monitoring the flood flow, assessment of the flood inundation areas and related damages, post-flood
 271 landscape changes, and pre-flood prediction are therefore seriously required. Among various scales of
 272 approaches for flood hazards (Sohn et al., 2008), the RPAS has been adopted for each purpose of the flood
 273 damage prevention and mitigation because it has an ability of quick measurement at a low cost (DeBell et
 274 al., 2016; Nakamura et al., 2017). Figure 5 shows an example of the use of RPAS for prompt damage
 275 assessment by a severe flood occurred on early July 2017 at northern Kyushu area, southwest Japan. The
 276 Geospatial Information Authority of Japan (GSI) utilized an RPAS for the post-flood video recording and
 277 photogrammetric mapping of the damaged area with flood flow and large woody debris.



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279 Figure 5. Image captures of flood hazard using RPAS just after the 2017 Northern Kyushu Heavy Rain in the
 280 early July (southwest Japan), provided by GSI. (a) A screenshot of the aerial video of a flooded area along
 281 the Akatani River, Asakura City in Fukuoka Prefecture. (b) Orthorectified image of the damaged area.
 282 Locations of woody debris jam are mapped and shown on the online map (GSI, 2017). The video and map
 283 products are freely provided (compatible with Creative Commons Attribution 4.0 International).

284

285 2.2.1. Potential analysis of flood inundation

286 The risk assessments of flood inundation before the occurrence of a flood is crucial for the mitigation of the
 287 flood-disaster damages. RPAS is capable of providing quick and detailed analysis of the land surface
 288 information including topographic, land cover, and land use data, which are often incorporated into the
 289 hydrological modelling for the flood estimate (Costa et al., 2016). As a pre-flood assessment, Li et al. (2012)
 290 explored the area around an earthquake-derived barrier lake using an integrated approach of remote sensing
 291 including RPAS for the hydrological analysis of the potential dam-break flood. They proposed a technical
 292 framework for the real-time evacuation planning by accurately identifying the source water area of the
 293 dammed lake using a RPAS, followed by along-river hydrological computations of inundation potential.
 294 Tokarczyk et al. (2015) showed that the UAVRPAS-derived imagery is useful for the rainfall-runoff modelling
 295 for the risk assessment of floods by mapping detailed land-use information. As a key input data, high-
 296 resolution imperviousness maps were generated for urban areas from UAVRPAS imagery, which improved
 297 the hydrological modelling for the flood assessment. Zazo et al. (2015) and Şerban et al. (2016) demonstrated
 298 hydrological calculations of the potentially flood-prone areas using UAVRPAS-derived 3D models. They
 299 utilized 2D cross profiles derived from the 3D model for the hydrological modelling.

300

301 2.2.2. Flood monitoring

302 Monitoring of the ongoing flood is potentially important for the real-time evacuation planning. Le Coz et al.
 303 (2016) mentioned that the movies captured by a RPAS, which can be operated by not only research specialists
 304 but also general non-specialists, is potentially useful for the quantitative monitoring of floods including flow
 305 velocity estimate and flood modelling. This can also contribute to the crowdsourced data collection for flood
 306 hydrology as the citizen science. In case of flood monitoring, however, areas under water is often problematic
 307 by image-based photogrammetry because the bed is not often fully seen in aerial images. If the water is clear
 308 enough, bed images under water can be captured, and the bed morphology can be measured with additional
 309 corrections of refraction (Tamminga et al., 2015; Woodget et al., 2015), but the flood water is often unclear
 310 because of the abundant suspended sediment and disturbing flow current. Another option is the fusion of

311 different datasets using a sonar-based measurement for the water-covered area, which is registered with
312 the terrestrial datasets (Flener et al., 2013; Javernick et al., 2014). Image-based topographic data of water
313 bottom by unmanned underwater vehicle (UUV, also known as an autonomous underwater vehicle, AUV)
314 can also be another option (e.g., Pyo et al., 2015), although such the application of UUV to flooding has been
315 limited.

316 Not only the use of topographic datasets derived from [Structure from Motion-Multi Stereo View \(SfM-MVS\)](#)
317 photogrammetry, the use of orthorectified images concurrently derived from the RPAS-based aerial images
318 is advantageous for the assessment of hydrological observation and modelling of floods. Witek et al. (2014)
319 developed an experimental system to monitor the stream flow in real time for the prediction of overbank
320 flood inundation. The real-time prediction results are also visualized online with a web map service with a
321 high-resolution image (3 cm/pix). Feng et al. (2015) reported that the accurate identification of inundated
322 areas is feasible using [UAVRPAS](#)-derived images. In their case, deep learning approaches of the image
323 classification using optical images and texture by [UAVRPAS](#) successfully extracted the inundated areas, which
324 must be useful for flood monitoring. Erdelj et al. (2017) proposed a system that incorporates multiple RPAS
325 devices with wireless sensor networks to perform the real-time assessment of a flood disaster. They
326 discussed the technical strategies for the real-time flood disaster management including the detection,
327 localization, segmentation, and size evaluation of flooded areas from RPAS-derived aerial images.

328

329 2.2.3. Post-flood changes

330 Post-flood assessments of the land surface materials including topography, sediment, and vegetation are
331 more feasible by RPAS surveys. Smith et al. (2014) proposed a methodological framework for the immediate
332 assessment of flood magnitude and affected landforms by SfM-MVS photogrammetry using both aerial and
333 ground-based photographs. In this case, it is recommended to carefully select appropriate platforms for SfM-
334 MVS photogrammetry (either airborne or ground-based) based on the field conditions. Tamminga et al.
335 (2015) examined the 3D changes in river morphology by an extreme flood event, revealing that the changes
336 in reach-scale channel patterns of erosion and deposition are poorly modelled by the 2D hydrodynamics
337 based on the initial condition before the flood. They also demonstrate that the topographic condition can be
338 more stable after such an extreme flood event. Langhammer et al. (2017) proposed a method to
339 quantitatively evaluate the grain size distribution using optical images taken by a RPAS, which is applied to
340 the sediment structure before and after a flash flood.

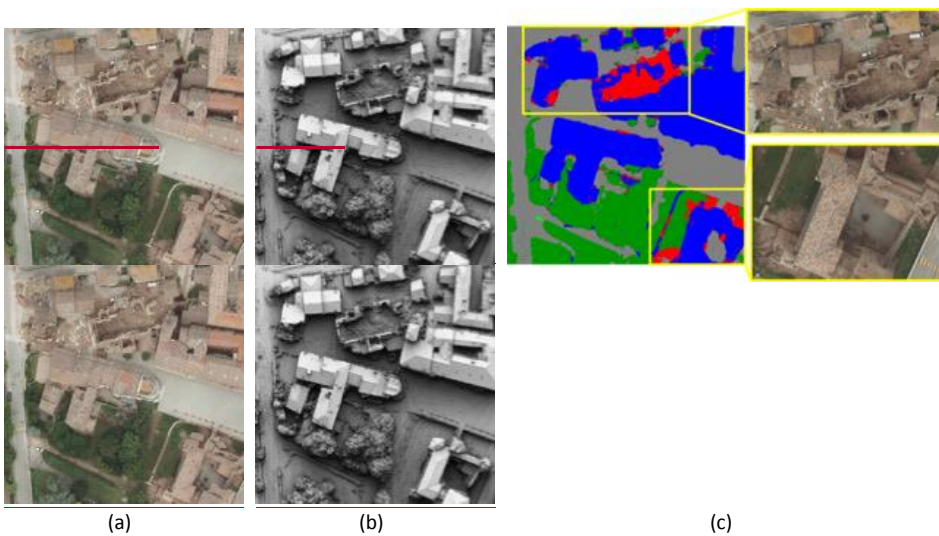
341 As a relatively long-term study, Dunford et al. (2009) and Hervouet et al. (2011) explored annual landscape
342 changes after the flood using RPAS-derived images together with other datasets such as satellite image
343 archives or a manned motor paraglider. Their work assessed the progressive development of vegetation on
344 a braided channel at an annual scale, which appears to be controlled by local climate including rainfall,
345 humidity, and air temperature, hydrology, groundwater level, topography, and seed availability. Changes in
346 the sediment characteristics by a flood is another key feature to be examined.

347

348 2.3 Earthquakes

349 Remote sensing technology has been recognized as a suitable source to provide timely data for automated
350 detection of damaged buildings for large areas (Dong and Shan, 2013; [Pham et al., 2014](#)). In the post-event,
351 satellite images have been traditionally used for decades to visually detect the damages on the buildings to

352 prioritize the interventions of rescuers. Operators search for externally visible damage evidence such as
353 spalling, debris, rubble piles and broken elements, which represent strong indicators of severe structural
354 damage. Several researches, however, have demonstrated how this kind of data often leads to the wrong
355 detection, usually underestimating the number of the collapsed building because of their reduced resolution
356 on the ground. In this regard, airborne images and in particular oblique acquisitions (Tu et al., 2017; Nex et
357 al., 2014; Gerke and Kerle 2011; [Nedjati et al., 2016](#)) have demonstrated to be a better input for reliable
358 assessments, allowing the development of automated algorithms for this task (Figure 6). The deployment of
359 photogrammetric aeroplanes on the strike area is however very often unfeasible especially when the early
360 (in the immediate hours after the event) damage assessment for response action is needed.



361 Figure 6. True-orthophoto, Digital Surface Model and damage map of an urban area using airborne nadir
362 images (Source: Nex et al., 2014).

363 For this reason, RPASs have turned out to be valuable instruments for the building damage assessment.
364 ([Hirose et al., 2015](#)). The main advantages of RPASs are their availability (and reduced cost) and the ease to
365 repeatedly acquire high-resolution images. Thanks to their high resolution, their use is not only limited to the
366 early impact assessment for supporting rescue operations, but it is also considered in the preliminary analysis
367 of the structural damage assessment.

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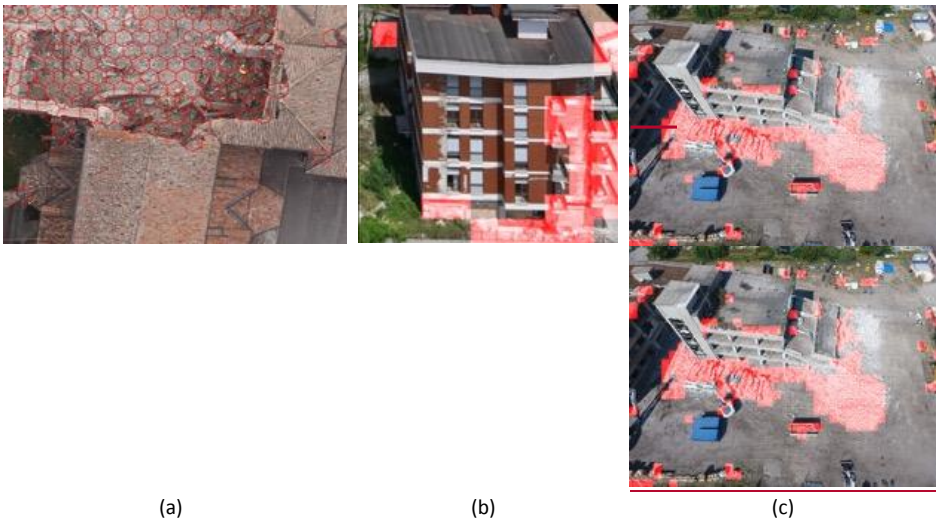
369 2.3.1 Early impact assessment

370 The fast deployment in the field, the easiness of use and the capability to provide in real time high-resolution
371 information of inaccessible areas to prioritize the operator's activities are the strongest point of RPASs for
372 these activities (Boccardo et al., 2015). The use of RPASs for rescue operations started almost a decade ago
373 (Bendea et al., 2008) but their massive adoption has begun only in the very last few years (Earthquake in
374 Nepal 2015) thanks to the development of low cost and easy to use platforms. Initiatives like- *UAViators*
375 (<http://uaviators.org/>) have further increased the public awareness and acceptance of this kind of
376 instruments. Several rescue departments have now introduced [small UAVs RPAS](#) as part of the conventional

377 equipment of their teams. (Xie et al., 2014). The huge number of videos acquired by UAVsRPAS and posted
378 by rescuers online (i.e. Youtube) after the 2016 Italian earthquakes confirm this general trend.

379 The operators use RPASs to fly over the interest area and get information through visual assessment of the
380 streaming videos. The quality of this analysis is therefore limited to the ability of the operator to fly the RPAS
381 over the interest area. The lack of video geo-referencing usually reduces the interpretability of the scene and
382 the accurate localization of the collapsed parts: only small regions can be acquired in a single flight. The lack
383 of georeferenced maps prevents the smooth sharing of the collected information with other rescue teams
384 limiting the practical exploitation of these instruments. UAVsRPASs are mainly used in daylight conditions as
385 the flight during the night is extremely critical, and the use of thermal images is of limited help for the
386 rescuers.

387 Many researchers have developed algorithms to automatically extract damage information from imagery
388 (Figure 7). The main focus of these works is to reliably detect damages in a reduced time to satisfy the time
389 constraints of the rescuers. In (Vetrivel et al., 2015) the combined use of images and photogrammetric point
390 clouds have shown promising results thanks to a supervised approach. This work, however, highlighted how
391 the classifier and the designed 2D and 3D features were hardly transferable to different datasets: each scene
392 needed to be trained independently strongly limiting the efficiency of this approach. In this regard, the recent
393 developments in machine learning (i.e. Convolutional Neural Networks, CNN) have overcome these limits
394 (Vetrivel et al., 2017, in press), showing how they can correctly classify scenes even if they were trained using
395 other datasets: a trained classifier can be directly used by rescuers on the acquired images without need for
396 further operations. The drawback of these techniques is the computational time: the use of CNN, processing
397 like image segmentation or point cloud generation are computationally demanding and hardly compatible
398 with real-time needs. In this regard, most recent solutions exploit only images (i.e. no need to generate point
399 cloud) and limit the use of most expensive processes to the regions where faster classification approaches
400 provide uncertain results to deliver an almost real-time information (Duarte et al., 2017).



401 Figure 7. Examples of damage detection on images acquired in three different scenarios (a) Mirabello (source: Vetrivel
402 et al., 2017, in press) and (b) L'Aquila and Lyon (source Duarte et al., 2017).

403

404 2.3.2 Building damage assessment

405 The damage evidence that can be captured from a UAV is not sufficient to infer the actual damage state of
406 the building as it requires additional information such as damages to internal building elements (e.g., columns
407 and beams) that cannot be directly defined from images. Even though this information is limited, images can
408 provide useful information about the external condition of the structure, evidencing anomalies and damages
409 and providing a first important information for structural engineers. Two main typologies of investigations
410 can be performed: (i) the use of images for the detection of cracks or damages on the external surfaces of
411 the building (i.e. walls and roofs) and (ii) the use of point clouds (generated by photogrammetric approach)
412 to detect structural anomalies like tilted or deformed surfaces. In both cases, the automated processing can
413 only support and ease the work of the expert who still interprets and assess the structural integrity of the
414 building.

415 In (Fernandez-Galarreta et al., 2015) a comprehensive analysis of both point clouds and images to support
416 the ambiguous classification of damages and their use for damage score was presented. In this paper, the
417 use of point clouds was considered efficient for more serious damages (partial or complete collapse of the
418 building), while images were used to identify smaller damages like cracks that can be used as the basis for
419 the structural engineering analysis. The use of point clouds is investigated in (Baiocchi et al., 2013; Dominici
420 et al., 2017): this contribution highlights how point clouds from UAVs can provide very useful information to
421 detect asymmetries and small deformations of the structure.

422

423 2.4 Volcanic activity

424

425 RPAS is particularly advantageous when the target area of measurement is hardly accessible on the ground
426 due to dangers of volcanic gas or risks of eruption in volcanic areas (Andrews, 2015). Although an equipment
427 of RPAS can be lost or damaged by the volcanic activities, the operator can safely stay in a remote place.
428 Various sensors can be mounted on a RPAS to monitor volcanic activities including topography, land cover,
429 heat, gas composition, and even gravity field (Saiki and Ohba, 2010; Deurloo et al., 2011, 2012; Astuti et al.,
430 2009; Middlemiss et al., 2016). The photogrammetric approach to obtain topographic data is widely applied
431 because RGB camera sensors are small enough to be mounted on a small aircraft. LargerAs mentioned
432 before, this paper considers in particular small RPAS. In the study of volcanoes, larger aircrafts with a payload
433 of kilograms are also utilized to mount other types of sensors to monitor various aspects of their dynamic
434 volcanic activities. For this reason, in this chapter we consider also larger RPAS solutions.

435

436 2.4.1. Topographic measurements of volcanoes

437 Long-distance flight of a RPAS enables quick and safe measurements of an emerging volcanic island. Tobita
438 et al. (2014a) successfully performed a fixed-wing RPAS flight for a one-way distance of 130 km in total flight
439 time of 2 hours and 51 minutes over the sea to capture aerial images of a newly formed volcanic island next
440 to Nishinoshima Island (Ogasawara Islands, southwest Pacific). They performed SfM-MVS photogrammetry
441 of the aerial images taken back from the RPAS to generate a 2.5 m resolution DEM of the island. The team
442 also performed two successive measurements of Nishinoshima Island in the following 104 days, revealing the

Commentato [G11]: AC8

443 morphological changes in the new island covering a 1,600 m by 1,400 m area (Nakano et al., 2014; Tobita et
444 al., 2014b).

445 Since the volcanic activities often last for a long period, it is also important to connect the recent volcanic
446 morphological changes to those in the past. Although detailed morphological data of volcanic topography is
447 often unavailable, historical aerial photographs taken in the past decades can be utilized to generate
448 topographic models at a certain resolution. Some case studies have used archival aerial photographs in
449 volcanoes for periods of more than 60 years, generating DEMs with resolutions of several meters for areas
450 of 10 km² (Gomez, 2014; ~~Darrien~~[Derrien](#) et al., 2015; Gomez et al. 2015). Although these DEMs are coarser
451 than those derived from RPAS, they can be used as supportive datasets for the modern morphological
452 monitoring using RPAS at a higher resolution and measurement frequency.

453

454 **2.4.2. Gas monitoring and product sampling**

455 Caltabiano et al. (2005) proposed the architecture of a RPAS for the direct monitoring of gas composition in
456 volcanic clouds of Mt. Etna in Italy. In this system, the 2-m wide fixed-wing RPAS can fly autonomously up to
457 4000 m altitude with a speed of 40 km/h. Like this system, a RPAS with a payload of several kilograms can
458 carry multiple sensors to monitor different compositions of volcanic gas. McGonigle et al. (2008) used a RPAS
459 for volcanic gas measurements at La Fossa crater of Mt. Vulcano in Italy. The RPAS has 3 kg payload and
460 allows to host an ultraviolet spectrometer, an infrared spectrometer, and an electrochemical sensor on
461 board. The combination of these sensors enabled the estimation of the flux of SO₂ and CO₂, which are crucial
462 for revealing the geochemical condition of erupting volcanoes. The monitoring of gas composition including
463 CO₂, SO₂, H₂S, H₂, as well as the air temperature, can be used for the quantification of the degassing activities
464 and prediction of the conduit magma convection, as suggested by the ~~test~~[tests](#) at ~~Mt. Kirishima~~[several](#)
465 [volcanoes](#) in Japan (Shinohara, 2013; [Mori et al., 2014](#)) and in Costa Rica ([Diaz et al., 2015](#)).

466 A RPAS can also transport a small ground-running robot (Unmanned Ground Vehicle: UGV) to slope head of
467 an active volcano, where the UGV takes close-range photographs of volcanic ash on the ground surface by
468 running down the slope (Nagatani et al., 2013). Protocols for direct sampling of volcanic products using a
469 RPAS have also been developed (Yajima et al., 2014).

470

471 **2.4.3. Geothermal monitoring**

472 In New Zealand, Harvey et al. (2016) and Nishar et al. (2016) carried out experimental studies on the regular
473 monitoring of intense geothermal environments using a small RPAS. They used thermal images taken by an
474 infrared imaging sensor together with normal RGB images for photogrammetry, mapping both the ground
475 surface temperature with detailed topography and land cover data. Chio and Lin (2017) further assessed the
476 use of a RPAS equipped with a thermal infrared sensor for the high-resolution geothermal image mapping in
477 a volcanic area in Taiwan. They improved the measurement accuracies using an onboard sensor capable of
478 post-processed kinematic GNSS positioning. This allows accurate mapping with less ground control points,
479 which are hard to place on such intense geothermal fields.

480

481 2.5 Wildfires

482 Wildfires are a phenomenon with local and global effects (Filizzola et al., 2017). Wildfires represent a serious
483 threat for land managers and property owners; in the last few years, this threat has significantly expanded
484 (Peters et al., 2013). The literature also suggests that climate change will continue to enhance the potential
485 forest fire activity in different regions of the world (McKenzie et al. 2014; Abatzoglou and Williams, 2016).
486 Remote sensing technologies can be very useful in monitoring such hazard (Shroeder et al., 2016). Several
487 scientists in the last few years used satellites in fire monitoring (Shroeder et al., 2016). More recently UAVs,
488 RPASs have been considered to be useful as well- (Martinez-de Dios et al., 2011). Hinkley and Zajkowski
489 (2011) presented the results of a collaborative partnership between NASA, and the US Forest Service
490 established for testing thermal image data for wildfires monitoring. A small unmanned airborne system
491 served as a sensor platform. The outcome was an improved tool for wildfire decision support systems. Merino
492 et al. (2012) described a system for forest fire monitoring using a UASRPAS. The system integrates the
493 information from the fleet of different vehicles to estimate the evolution of the forest fire in real time. The
494 field tests indicated that UASRPAS could be very helpful for the activities of firefighting (e.g. monitoring).
495 Indeed, they cover the gap between the spatial scales given by satellites and those based on cameras. Wing
496 et al. (2014) underlined the fact that spectral and thermal sensors mounted in UAVsRPASs may hold great
497 promise for future remote sensing applications related to forest fires. UASsRPASs have greater potential to
498 provide enhanced flexibility for positioning and repeated data collection. Tang and Shao (2015) summarize
499 various approaches of remote drone sensing to surveying forests, mapping canopy gaps, measuring forest
500 canopy height, tracking forest wildfires, and supporting intensive forest management. These authors
501 underlined the usefulness in using drones for wildfire monitoring. UAVsRPASs can repeatedly fly to record
502 the extent of an ongoing wildfire without jeopardizing crews' safety. Zajkowski et al. (2015) tested different
503 UAVsRPASs (e.g. quadcopter, single-fixed-wing) for the analysis of fire activity. Measurements included visible
504 and long-wave infrared (LWIR) imagery, black carbon, air temperature, relative humidity and three-
505 dimensional wind speed and direction. The authors also described in detail the mission's plan, including the
506 logistics of integrating RPAS into a complex operations environment, specifications of the aircraft and their
507 measurements, execution of the missions and considerations for future missions. Allison et al. (2016)
508 provided a detailed state of the art on fire detection using both manned and unmanned aerial platforms. This
509 review highlighted the following challenges: the need to development of robust automatic detection
510 algorithms, the integration of sensors of varying capabilities and modalities, the development of best
511 practices for the use of new sensor platforms (e.g. small UAVsmini RPAS), and their safe and effective
512 operation in the airspace around a fire.

513

514 3. Discussion and conclusion

515 In this paper, we analysed possible applications of RPAS to natural hazards. The available literature on this
516 topic is strongly increased in last few years, according to the improvement of the diffusion of these systems.
517 In particular, we considered: landslides, floods, earthquakes, volcanic activities and wildfires.

518 RPAS can support studies on active geological processes and can be considered a good solution for the
519 identification of effects and damages due to several catastrophic events. One of the most important elements
520 that characterized the use of RPAS is their flexibility and versatility, largely confirmed by the wide number of
521 operative solutions available in the literature. The available literature pointed out the necessity of the
522 development of dedicated methodologies that can be able to take the full advantage of RPAS. In particular,
523 typical results of structure from motion software (orthophoto and DSM) that are considered the end of

524 standard data-processing, can be very often the starting point of dedicated procedures specifically conceived
525 for natural hazards applications.

526 In the pre-emergency phase, one of the main advantages of RPAS surveys is to acquire high resolution and
527 low-cost data to analyse and interpret environmental characteristics and potential triggering factors (e.g.
528 slope, lithology, geostructure, land use/land cover, rock anomalies, and displacement). The data can be
529 collected with high revisit times to obtain multi-temporal observations. After the characterization of hazard
530 potential and vulnerability, some areas can be identified by a higher level of risk. These cases request an
531 intensive monitoring, to gain a quantitative evaluation of the potential occurrence of an event. In this
532 context, the use of aerial data represents a very useful complementary data source concerning the
533 information acquired through ground-based observations in particular for dangerous areas.

534 During the emergency phase, high-resolution imagery is asked to be acquired over the event site. The primary
535 use of this data is for the assessment of the damage grade (extent, type and damage grades specific to the
536 event and eventually of its evolution). They may also provide relevant information that is specific to critical
537 infrastructures, transport systems, aid and reconstruction logistics, government and community buildings,
538 hazard exposure, displaced population, etc. (Ezequiel et al., 2014). Concurrently, the availability of clear and
539 straightforward raster and vector data, integrated with base cartographic contents (transportation, surface
540 hydrology, boundaries, etc.) it is recognized as an added-value to support decision makers for the
541 management of emergency operations. (Fikar et al., 2016). These applications very often need prompt and
542 reliable interventions. RPAS should, therefore, deliver information promptly. In this regard, very few
543 researchers have focused on this issue: most of the reported workworks present (often time-consuming and
544 even manual) post-processing of the acquired data, precluding the use of their results from practical and
545 real-life scenarios. A big effort should be taken by the research community to propose faster and automated
546 approaches. In particular during emergencies, the time required for RPAS dataset processing is an important
547 element that should be carefully considered. Giordan et al. (2015a) presented a case study related to a
548 landslide emergency. In this paper, authors considered not only possible results but also the time that is
549 required for them

550 As in many other domains, RPAS present a disruptive technology where, beside conventional SfM
551 applications for 3D reconstructions, many dedicated and advanced methodologies are still in their
552 experimental phase and will need to be further developed in the incoming years. In the following years, it
553 would be desirable to witness the transfer of the best practices in the use of RPAS be then from the Research
554 community to Government Agencies (or private companies) involved in the prevention and reduction of
555 impacts of natural hazards. The Scientific community should contribute to the definition of standard
556 methodologies that can be assumed by civil protection agencies for the management of emergencies.

Commentato [G12]: AC14

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