

1 Slope stability and rock fall assessment of volcanic tuffs using RPAS 2 with 2D FEM slope modelling

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9 **Abstract.** Steep, hardly accessible cliffs of rhyolite tuff in NE Hungary are prone to rock falls endangering visitors of a
10 castle. Remote sensing techniques were employed to obtain data on terrain morphology and to provide slope geometry for
11 assessing the stability of these rock walls. RPAS (Remotely Piloted Aircraft System) was used to collect images which were
12 processed by Pix4D mapper (Structure-from-Motion technology) to generate a point cloud and mesh. The georeferencing
13 was made by Global Navigation Satellite System (GNSS) with the use of 7 ground control points. The obtained digital
14 surface model (DSM) was processed (vegetation removal) and the derived Digital Terrain Model (DTM) allowed to draw
15 cross sections and to detect joint system. Joint and discontinuity system was also verified by field measurements. On site as
16 well as laboratory tests provided additional engineering geological data for slope modelling. Stability of cliffs was assessed
17 by 2D FEM (Finite Element Method). Global analyses of cross-sections show that weak intercalating tuff layers may serve
18 as potential slip surfaces, however, at present; the highest hazard is related to planar failure along ENE-WSW joints and to
19 wedge failure. The paper demonstrates that RPAS is a rapid and useful tool of generating reliable terrain model of hardly
20 accessible cliff faces. It also emphasizes the efficiency of RPAS in rock fall hazard assessment in comparison with other
21 remote sensing techniques such as terrestrial laser scanning (TLS).

22 1 Introduction

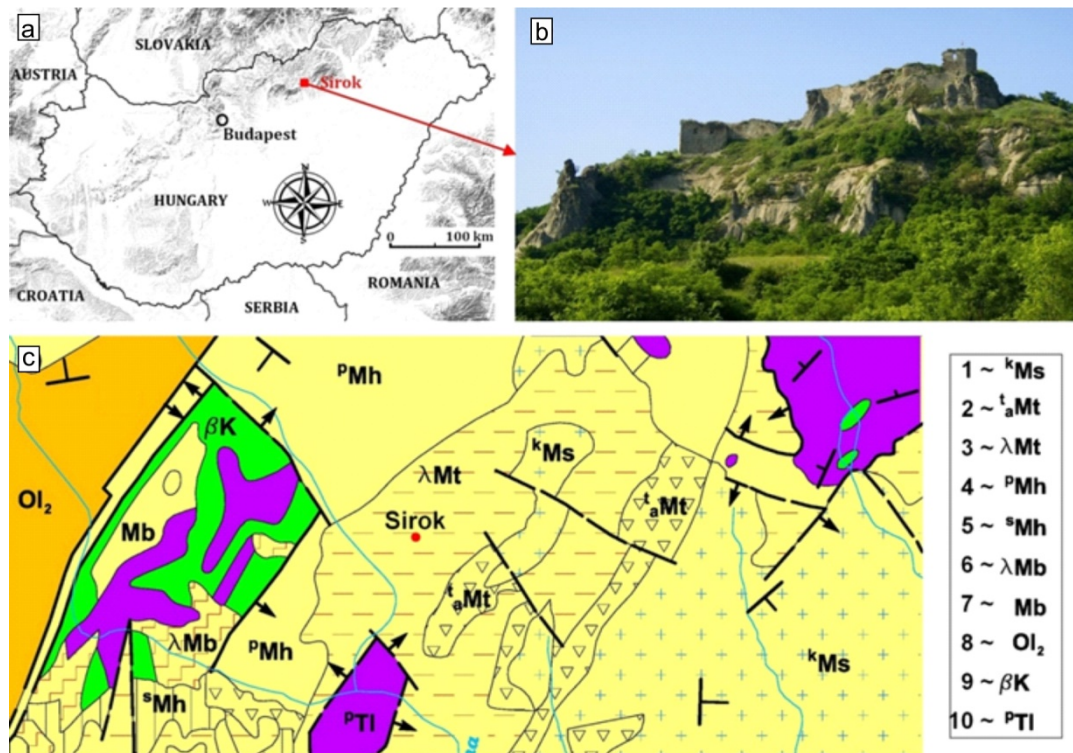
23 In the past years, technological development of RPAS revolutionized the data gathering of landslide affected areas (Rau et
24 al. 2011), recultivated mines (Haas et al. 2016) and the monitoring of coastal processes (Casella et al. 2016), and levee
25 breaches (Brauneck et al. 2016) or road cuts (Mateos et al. 2016). RPAS has been increasingly used in engineering geology
26 in historical landslide mapping (Jovančević et al. 2016) and in slope stability analyses (Niethammer et al. 2012, Fraštia et al
27 2014) as well as in other natural disasters such as earthquakes (Gerke & Kerle 2011, Nex et al. 2014) or floods (Feng et al.
28 2015). RPAS can also be combined with terrestrial laser scanning (TLS) since both remote sensing tools provide high
29 precision terrain measurement (Fanti et al. 2013, Assali et al. 2014, Francioni et al. 2014, Neugirg et al. 2016, Manconi &
30 Giordan 2015). These tools can be used to validate height information derived by other technologies. Rock falls represent
31 special landslide hazards since their rapid movements and various trajectories make it difficult to predict their hazard
32 potential (Crosta & Agliardi 2003, Manconi & Giordan 2014). Several methods have been suggested to assess cliff stability
33 ranging from physical prediction rock fall hazard index (Crosta & Agliardi 2003) via Rockfall Hazard Rating System
34 (Budetta 2004) to the modelling of their trajectories (Crosta & Agliardi 2002, Abbruzzese et al. 2009, Copons et al. 2009,
35 Samodra et al. 2016). These methods rely on understanding failure mechanisms and on predicting displacement of rock
36 masses (Pappalardo et al. 2014, Stead & Wolter 2015, Mateos et al. 2016), or in some cases, on predicting displacement of
37 individual rock blocks (Martino & Mazzanti 2014). To gather data on the rock fall hazard of existing cliff faces, a number of
38 crucial data is needed: slope profiles, material properties, block size (De Biagi et al. 2017), and possible discontinuity
39 surfaces that can contribute to slope instability. Slope profiles can be obtained from point clouds, while material properties

40 have to be measured on site (e.g. Uniaxial Compressive Strength by Schmidt hammer) or under laboratory conditions
 41 (Margottini et al. 2015). For the detection and mapping of joints and fractures it is possible to apply remote sensing
 42 techniques (Fanti et al. 2013, Tannant 2015, Salvini et al. 2017).
 43 Most rock fall hazard publications deal with hard, well cemented rocks, such as limestone (Samodra et al. 2016), or various
 44 other types of sedimentary rocks (Michoud et al. 2012) such as igneous or metamorphic rocks. In contrast, very few previous
 45 studies deal with cliff face stability and rock fall hazard of low strength rock such as volcanic tuffs (Fanti et al. 2013,
 46 Margottini et al. 2015). Volcanic tuffs are very porous rocks and are prone to weathering (Arikan et al. 2007). While the
 47 current paper deals with a low strength pyroclastic rock, it has a slightly different approach of cliff stability analysis. In this
 48 study, slope stability is assessed by using a combination of remote sensing techniques, field measurements, and laboratory
 49 testing of tuffs with 2D FEM (Finite Element Analysis) analyses of slopes. In contrast to other case studies, this study
 50 operates on a smaller scale and studies the possibilities of wedge and planar failures. More specifically, in our context, the
 51 cliff face is unstable as it is evidenced by falling blocks. Due to rock fall hazard, the small touristic pathway was closed to
 52 avoid casualties. The current paper analyses the cliff faces by condition assessment and stability calculations. Thus, this
 53 research provides an assessment of how RPAS-based images and photogrammetric processing can be used to derive a
 54 surface model at sites that are difficult to access. The paper also demonstrates the combined use of photogrammetric,
 55 surveying, and engineering geological methods at difficult ground conditions in assessing rock slope stability.

56 2 Study area

57 The study area is located at mid-mountain range in NE-Hungary, where a hardly accessible jointed rhyolite tuff cliff face
 58 was studied. On the top of the cliff, a touristic point, the Sirok Castle is located (Fig. 1), with the steep rhyolite tuff hill with
 59 an elevation of 298 m at the transition area of two mountain ranges, Mátra and Bükk Mountains. The tuff is very porous and
 60 prone to weathering (Vásárhelyi 2002, Kleb & Vásárhelyi 2003, Török et al. 2007).

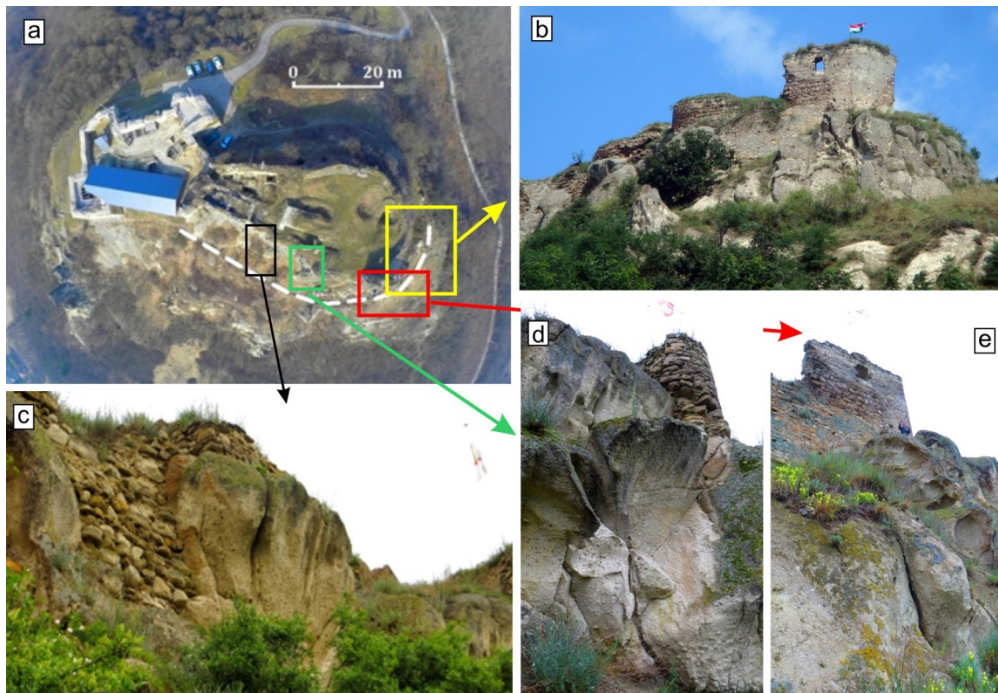
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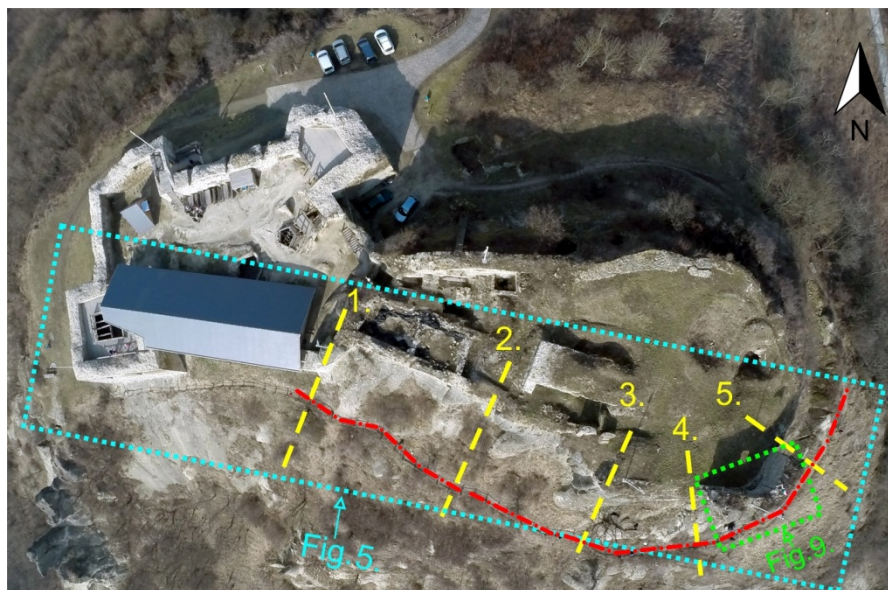
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63 Fig.1. Location of studied cliff faces and an image of the rocky slope at Sirok Castle, NE Hungary (top) and the geological map of
 64 the area (redrawn after Balogh 1964) (bottom). Legend for geological map: Miocene (1-7), Oligocene (8), Cretaceous (9), Triassic
 65 (10): 1: gravel and conglomerate; 2: clay; 3:rhyolite tuff; 4: sand and sandstone; 5: siltstone; 6: rhyodacite tuff; 7: fine sand; 8:
 66 clay; 9: basalt; 10: radiolarite.

67 Although the first castle was already constructed in the 13th century AD, due to war damages and reconstructions, the
 68 current structure encompasses wall sections representing different construction periods. In these days, the partially ruined
 69 walls have been restored, and the castle is open to tourists, but southern slopes are closed due to rock fall hazard.
 70 The hill represents a rhyolite tuff that was formed during the Miocene volcanism (Badenian-Lower Pannonian period). The
 71 cliff face was formed during to the late Miocene volcanic activity, and is a part of the Inner Carpathian volcanic chain. The
 72 geological map of the closer area clearly reflects the dominance of pyroclastic rocks, with isolated occurrences of Oligocene
 73 and Triassic rocks (Fig. 1). The cliffs are steep and display several joints and discontinuity surfaces. The present study
 74 focuses on the southern hillslope of the castle hill, where major rock falls occurred in the near past (Fig. 2). The study area is
 75 divided into smaller units, where RPAS and rock fall hazard assessment analyses were carried out (Fig 3).
 76



77
 78 **Fig.2. Studied southern cliff faces:** a) image of the castle obtained by RPAS with marked details; b) distant view of the eastern part
 79 of the cliff section; c) steep cliffs dissected by joints; d) vertical to sub-vertical cliff face with steep joints and traces of rock fall and
 80 e) weathered rounded cliff with larger taffoni.

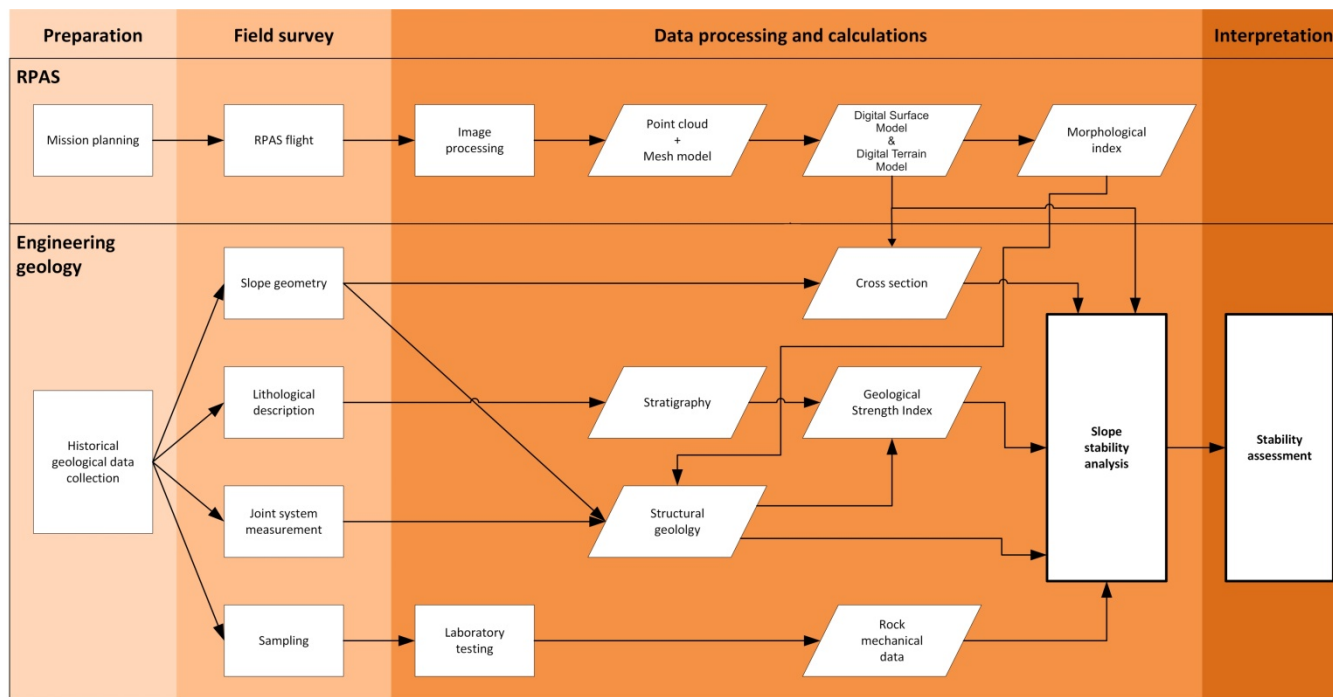


81
 82 **Fig.3. Location of the illustrations in the paper and the rock fall affected area.** Red dot-and-dash line represents zones affected by
 83 rock fall. Yellow dashed lines 1 to 5 mark the sections where slope stability was calculated by using 2D FEM model (Fig.8). Dotted
 84 lines indicate the areas shown on Fig. 5 and Fig. 9.

85 3 Materials and Methods

86 The research contains two major methods: i) RPAS and ii) engineering geology. The applied methods are summarized in a
 87 flow diagram displaying the combination and links between the two methods (Fig. 4). The flow chart has four major realms
 88 that have both RPAS and engineering geological units: i) preparation; ii) field survey; iii) data processing and calculation
 89 and finally interpretation. The RPAS line is described in details in the next part of the paper, but also linked to previous
 90 publications providing overview of image acquisition, image processing and interpretation (Civera et al. 2012, Westoby et al.
 91 2012, Remondino et al. 2014). Engineering geological part of the flow chart is also explained below, but also has strong
 92 links to publications describing the application of RPAS to landslide characterization, and rock slope stability assessment
 93 (Niethammer et al. 2012, Tannant 2015).

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97 **Fig.4. Flow chart showing the methods and obtained data set of this paper indicating the interrelationship between RPAS, slope**
 98 **stability assessment (see details in the text)**

99

100 3.1 RPAS data acquisition and terrain modelling

101 The Remotely Piloted Aerial System (RPAS) was deployed on 21st February 2015, when vegetation cover was limited. The
 102 remaining vegetation was manually removed; luckily, the areas with the highest hazard were barely covered. The system is a
 103 modified commercial DJI Phantom 2 drone (DJI, 2015), where the flying vehicle has been equipped with a synchronous
 104 image transfer (First-Person Viewer – FPV) option that also forwards the current flying parameters (e.g. height, speed, tilt,
 105 power reserve). Due to the complexity of the survey area, the flight was controlled manually; the required overlap between
 106 images was ensured by the operator considering capture frequency. The necessary overlap between images was controlled by
 107 the FPV option. For safety reasons, the crew consisted of two persons: one for controlling the aircraft, and the other one for
 108 continuously monitoring the transferred video stream. The camera control was done by a tablet.

109 A GoPro Hero 3+ (GoPro, 2017) action camera was mounted onto a 2-DoF gimbal of the unmanned aerial vehicle (UAV).
 110 The camera has a fixed 2.77 mm focal length objective that is capable of capturing 4000 × 3000 pixel sized JPG images. The

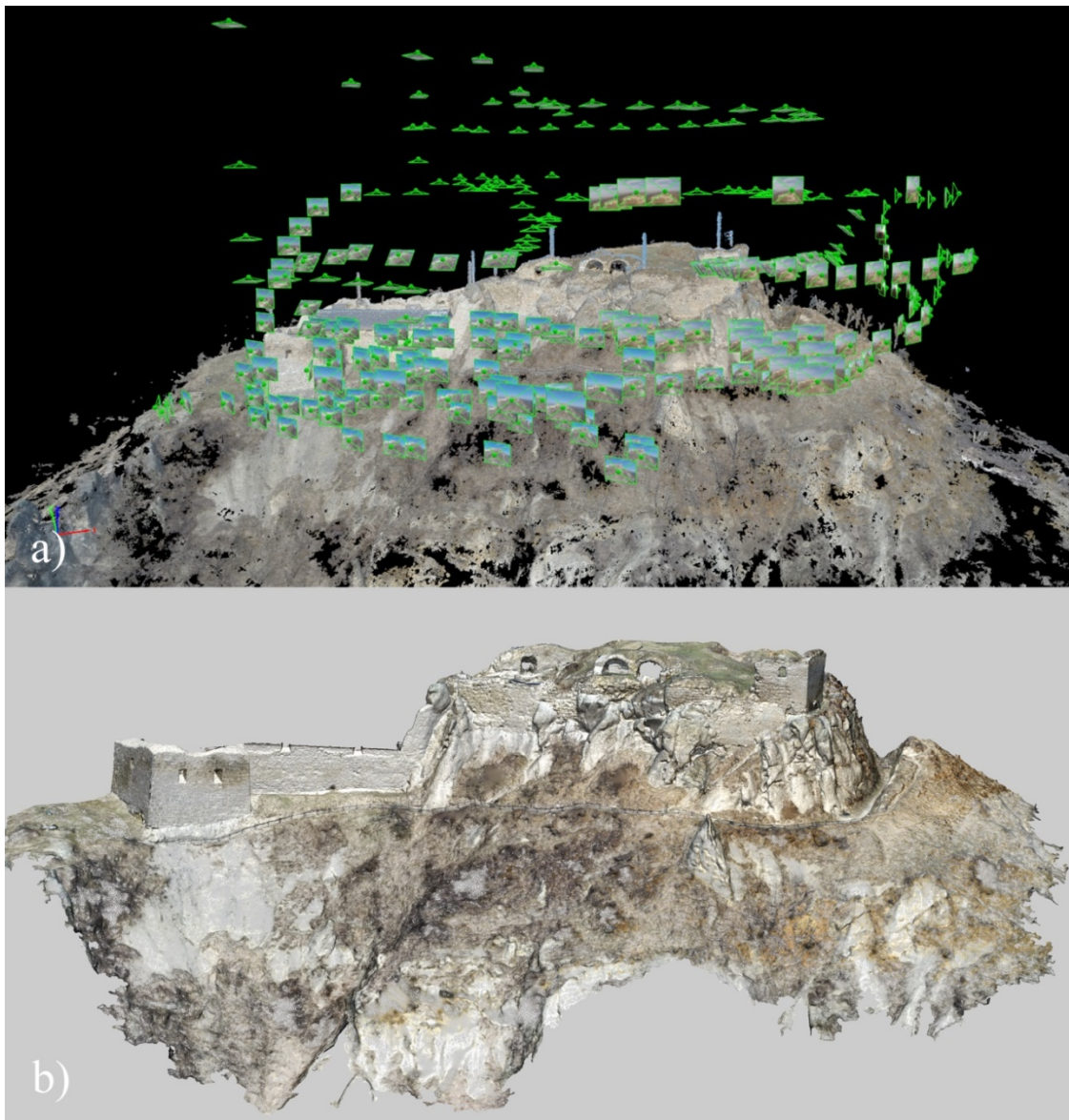
111 images were captured with a sensitivity of ISO 100 and RGB colour space. The lens was used with a fixed aperture of 2.8
112 and the camera was able to adjust the adequate shutter speed. Generally, the exposure time was set to 1/1400 s and the
113 images were compressed at a rate of 4.5 bits/pixel. There were three imaging flights; two around noon and one at about 5 in
114 the afternoon. The flying times were 13, 12 and 13 minutes, respectively, where 390, 365 and 419 images were captured. All
115 1174 images were involved in the photogrammetric object reconstruction (Fig. 5). The photogrammetric reconstruction has
116 been done by Pix4Dmapper (Pix4D, 2017), which is based on Structure-from-Motion (SfM) technology (Lowe 2004,
117 Westoby et al. 2012, Danzi et al. 2013.). SfM automatically identifies tie points considering initial requirements (e.g.
118 preliminary image centre positions, time stamps) (Table 1). Camera calibration was executed during post-processing, and no
119 prior calibration was needed (Pix4D, 2017). After the image alignment, the image projection centres and attitudes can be
120 observed (Fig. 5). 19.3 million points were obtained by the photogrammetric reconstruction, which was appropriate for the
121 engineering geological application, however the technology allows to obtain higher resolution, but it was not necessary. The
122 average point density is about 670 points/m³, but there are areas, where double point density.

123

124 **Table 1. Image processing data**

Mean number of keypoints per image	22676
Mean number of matched keypoints per image	9546
Mean reprojection error [pixel]	0.176
Time for SfM processing	40m:20s
Time for densification (point cloud)	05h:30m:24s

125



126

127 **Fig.5. The captured image positions around the reconstructed castle hill (top) and the point clouds obtained by RPAS technology**
 128 **(bottom) (see top view on Fig. 3.)**

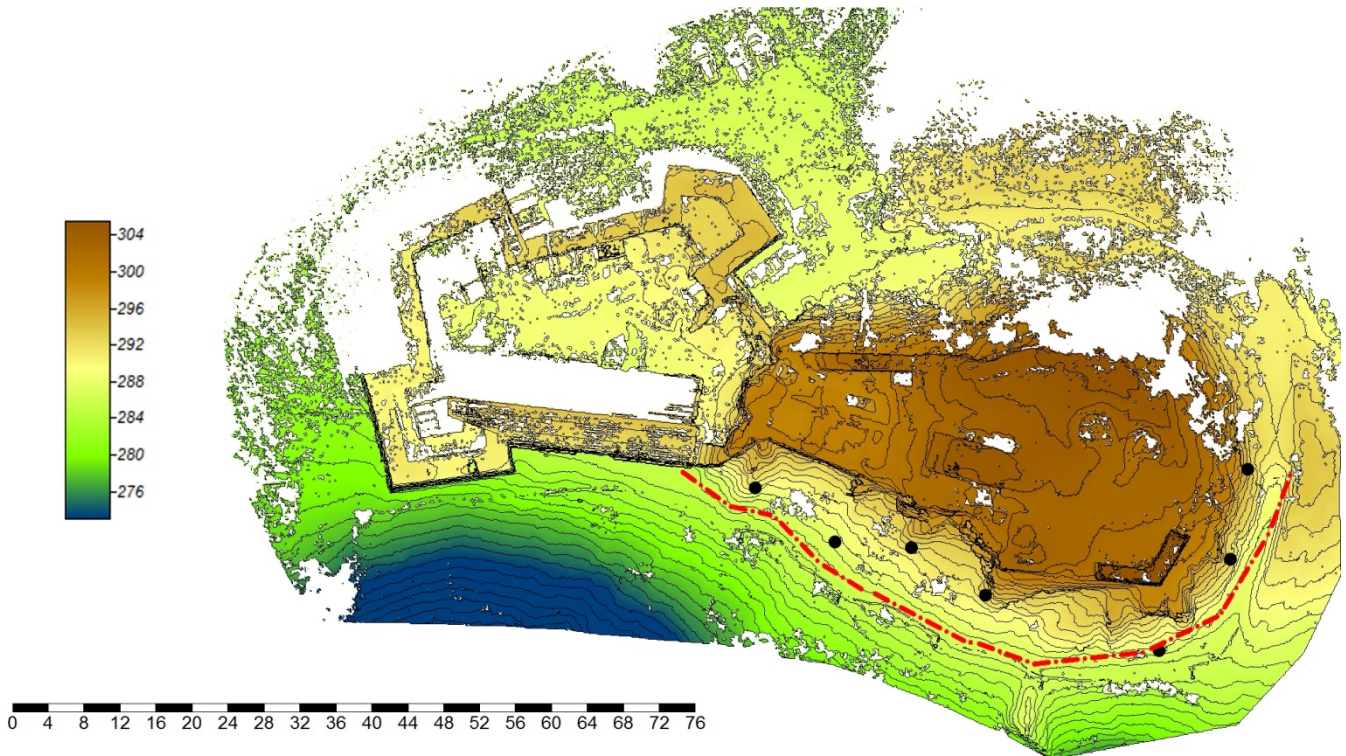
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130 For georeferencing, particular tie objects were measured by the Global Navigation Satellite System (GNSS). The used GNSS
 131 receiver was a Leica CS10 with a Gs08plus antenna (GS08, 2014, CS10, 2014). The measurement was done in RTK mode
 132 supported by the Hungarian RTK network (RTKnet, 2013). There were 7 measured ground control points (Fig. 6) (GCPs);
 133 the mean 3D measurement accuracy was 4.9 cm (minimal value was 2 cm, maximal value 9 cm). The RPAS technology has
 134 produced a considerable amount of data points (observations). Since this point cloud is difficult to manage due to its size,
 135 and heterogeneous point spacing, the latter processing requires a sophisticated resampling step, which was done by
 136 CloudCompare where the spatial resolution of the point cloud was set to 1 cm.

137 The RPAS data collection was validated by the use of terrestrial laser scanning. The necessary data were captured by two
 138 scanners: a Faro Focus S 120 3D (Faro, 2016) and a Z+F Imager 5010C (Z+F 2014). The terrestrial laser scanning was
 139 executed on the same day as the RPAS mission. The raw point cloud measured by Faro scanner contained 1.9 billion points,
 140 whilst the Z+F point cloud had 0.8 billion points. Both point clouds included X, Y, Z coordinates, intensity and RGB colour
 141 values. RPAS and TLS based point clouds were compared by CloudCompare software (CloudCompare, 2014) (Fig. 7).

142 As one can notice in Fig. 7, the highest difference between the two sources is almost less than 10 cm and the majority of the
 143 differences is about 1 cm. The point cloud was then imported into Geomagic Studio 2013 (GeomagicStudio, 2013) and

144 meshed, where the triangle side length was 5-7 cm. To support the engineering geological survey, several horizontal and
145 vertical sections were derived in Geomagic DesignX 2016 (GeomagicDesignX, 2016); these profiles were exported in CAD-
146 format.
147 The next step was to make cut-offs focusing only on the cliffs; it was done by CloudCompare, followed by the points being
148 exported in LAS-format (LAS, 2012). The exported points could then be imported into SAGA GIS 2.1.2 (Conrad et al.
149 2015), where the necessary DTMs were created by inverse distance weighting (IDW) algorithm (IDW, 2013). The derived
150 DSM-grids have 5 cm spatial resolution, which is adequate for morphologic analyses (Fig. 6) and suits to slope stability
151 analysis. The morphology analysis has concentrated on Catchment Area (CA) (Costa-Cabral and Burges 1994, Haas et al.
152 2016) (Fig 4), although several other morphological indices (e.g. Topographic Wetness Index, Stream Power Index) were
153 derived. These indices express the potential relationship between surface geometry and geological parameters.
154



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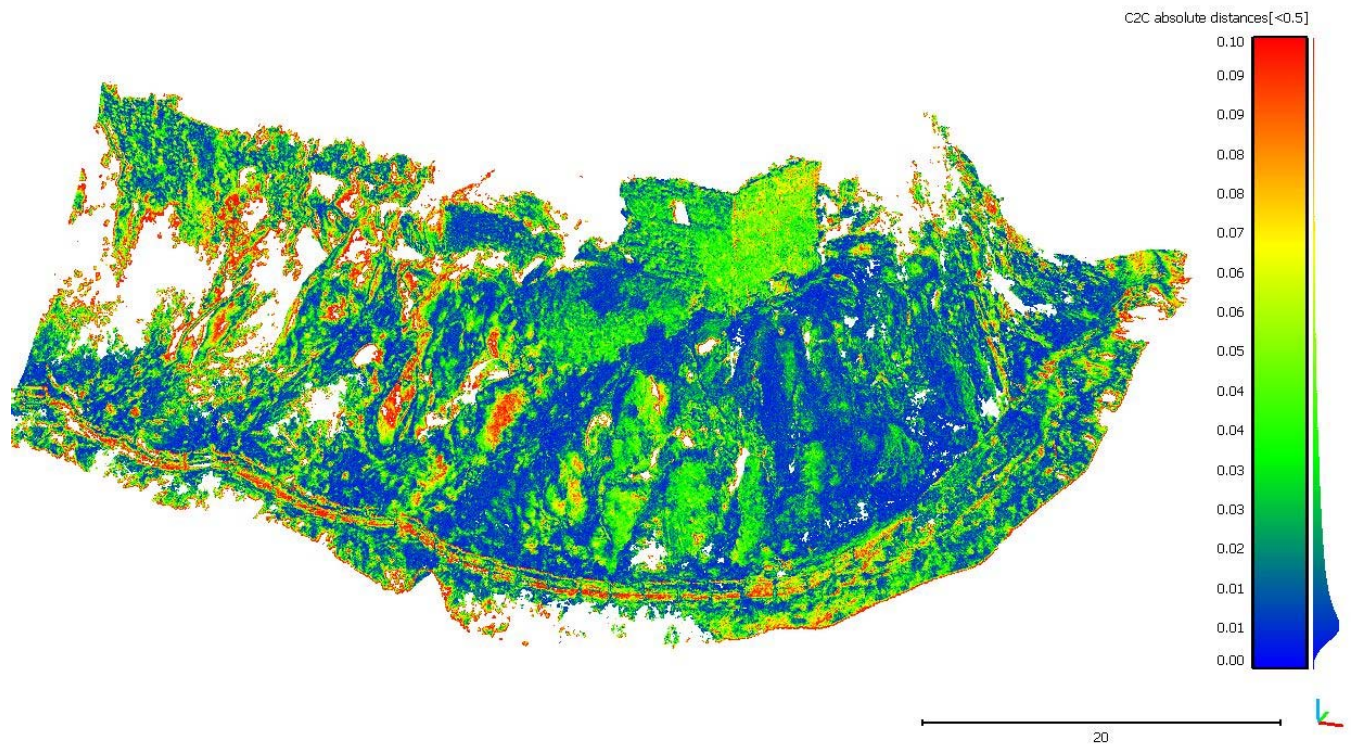
157 **Fig.6. Colour Digital Surface Model with 1 m contour line interval of the study area. The solid black dots show the ground control**
158 **points, while red dot-and-dash line represent zones affected by rock fall.**

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163
 164 **Fig. 7. Differences between RPAS and TLS point clouds by CloudCompare shown in metres (modus of differences is at about 0.01**
 165 **m)**

166

167 3.2 Engineering geology and slope stability analysis

168 Geological data and written resources (Balogh 1964, Haas 2013, Lukács et al. 2015) provided background information for
 169 the planning of engineering geological field survey (Fig. 4). Major lithotypes were identified and described and geological
 170 profiles were recorded during the engineering geological field surveys (Fig. 4). Rock joints, discontinuity surfaces, and fault
 171 systems were measured by using compass and structural geological software applied in mobile phone. The structural
 172 geological data was analysed by Dips software. Strength parameters were assessed on-site by using a Schmidt hammer. 10
 173 rebound values were measured on each surface and mean values and standard deviations were also calculated. This method
 174 has been used previously to gather rapid data on rock strength of cliff faces (Margottini et al. 2015). The data-set was
 175 compared to rock mechanical laboratory tests.

176 Samples for laboratory analyses were collected on site (Fig. 4). Major rock mechanical parameters were measured under
 177 laboratory conditions on cylindrical specimens. These were drilled from blocks and cut into by appropriate size using cutting
 178 disc. The sizes of tested specimen were made according to EN on air dry and on water saturated samples. The specimens
 179 were grouped according to the bulk density and the propagation speed of the ultrasonic pulse wave. Strength parameters such
 180 as uniaxial compressive strength, indirect tensile strength (Brazilian), were measured according to relevant EN standards and
 181 ISRM suggested methods. Modulus of elasticity was also calculated (Table 2). The generalized Hoek-Brown failure criterion
 182 (Hoek et al. 2002) was used to determine strength parameters of the rock mass. Altogether, 53 cylindrical test specimens
 183 were used for the tests.

184 The falling blocks can endanger the touristic footpath bellow the castle on the southern slopes; therefore, the stability
 185 analysis of the rocky slopes was focused on this part of the cliff (Fig. 3). First, the rock mass failure was analysed with by
 186 the RocFall FEM software of the Rocscience (RS2). The steepest sections were determined from the terrain model obtained
 187 from RPAS data. The GSI values of the rock masses were determined according to Marinós et al. (2005). The global stability
 188 of the hillslope of selected sections was calculated with RS2 software. Since the rhyolite tuff is a weak rock with few joints,

189 the rock mass failure and the failure along discontinuities were also analysed. This kinematic analysis had been done with a
 190 stereographic tool. The orientations of main joint sets were obtained from DSM model based morphological analyses with
 191 the use of Catchment Area tool. It assumes that major flow paths are related to joints, i.e. fracture system controls the
 192 drainage pattern (Costa-Cabral et al. 1994). At accessible areas joints and fractures were also measured on site on the
 193 southern and south-eastern parts of the hillslope. Additional control field measurements were also made in the underground
 194 cellar system of the castle, where the tuff is also exposed. The Dips software was used for the kinematic analysis. The
 195 direction of the hillslopes and the direction of the discontinuities were compared to determine the location of the potential
 196 hazardous failure zones on the hillside. Stereographic plots were generated showing the possible failure planes for all slope
 197 directions and the safety factor of the possible planar failure was calculated by Rockplane software. Wedge failure was
 198 modelled by Swedge software. Toppling failure due to geological and geomorphological conditions cannot occur. Slope
 199 stability calculations and stability assessment formed the last part of the engineering geological analyses (Fig. 4).

200

201 **Table 2. Rock mechanical tests and relevant standards.**

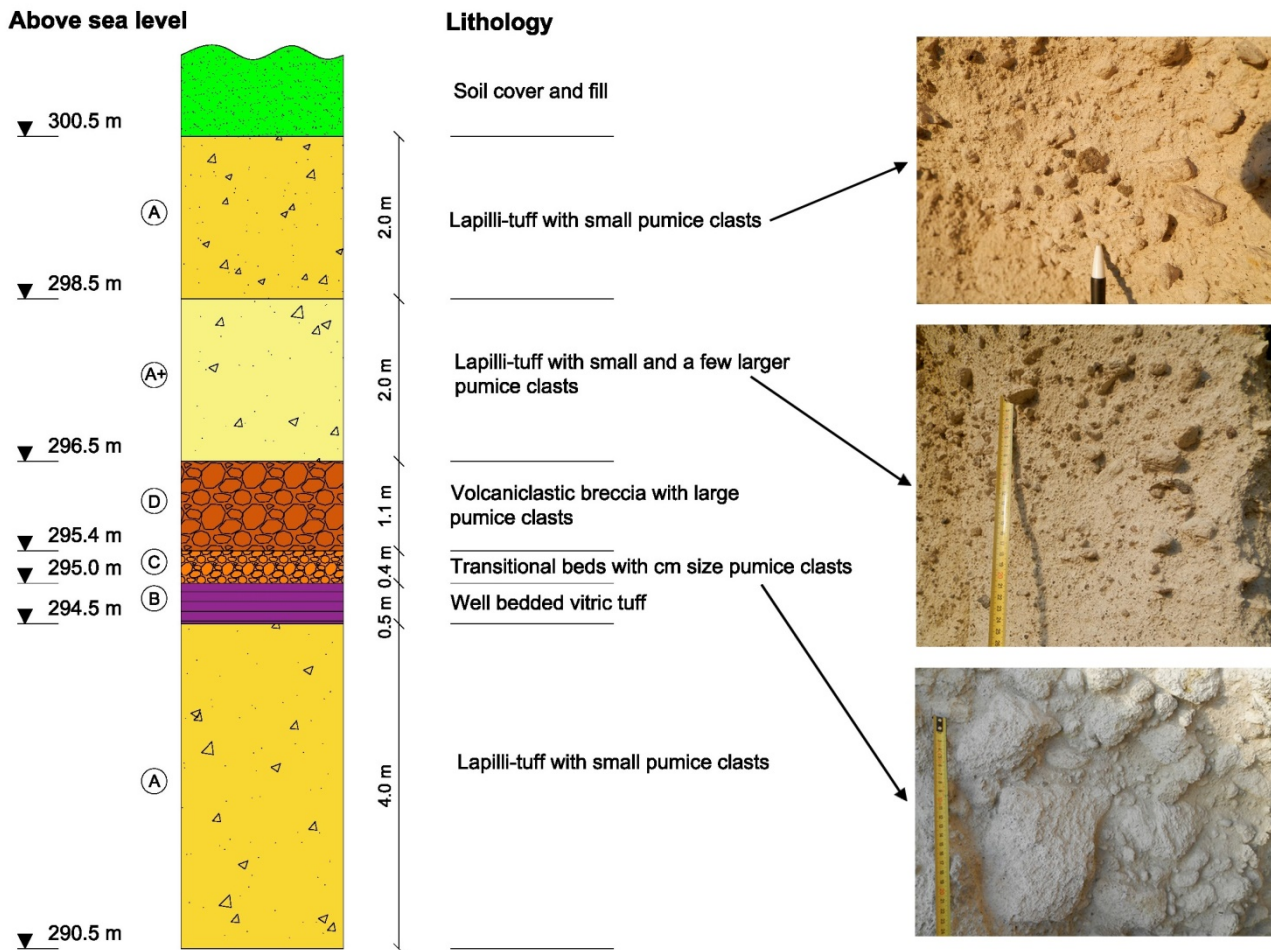
Rock mechanical parameter	Number of specimens	Relevant standard
Bulk density	53	EN 1936:2000
Water absorption	18	EN 13755:2008
Propagation speed of the ultrasonic wave	53	EN 14579:2005
Uniaxial compressive strength	31	ISRM 2015
Modulus of elasticity	31	ISRM 2015
Tensile strength (Brazilian)	23	ISRM 2015

202

203 **4. Results**

204 The rhyolite tuff faces consist of moderately bedded ignimbritic horizons, and also brecciated lapilli tuffs and tuffs according
 205 to our field observations (Fig. 6). The topmost 10 metres of the cliff face, which was modelled from slope stability,
 206 comprises 3 main horizons and can be modelled as “sandwich structure”. The lower and the upper parts are formed by thick
 207 pumice containing lapilli tuffs. These beds enclose nearly 2 metres of well-bedded less-welded fine tuff and brecciated
 208 horizons (Fig. 8).

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211

212 **Fig.8. Lithologic column of Sirok Várhegy showing the modelled topmost 10 metres section of the hill (letters refer to lithologic**
 213 **units)**

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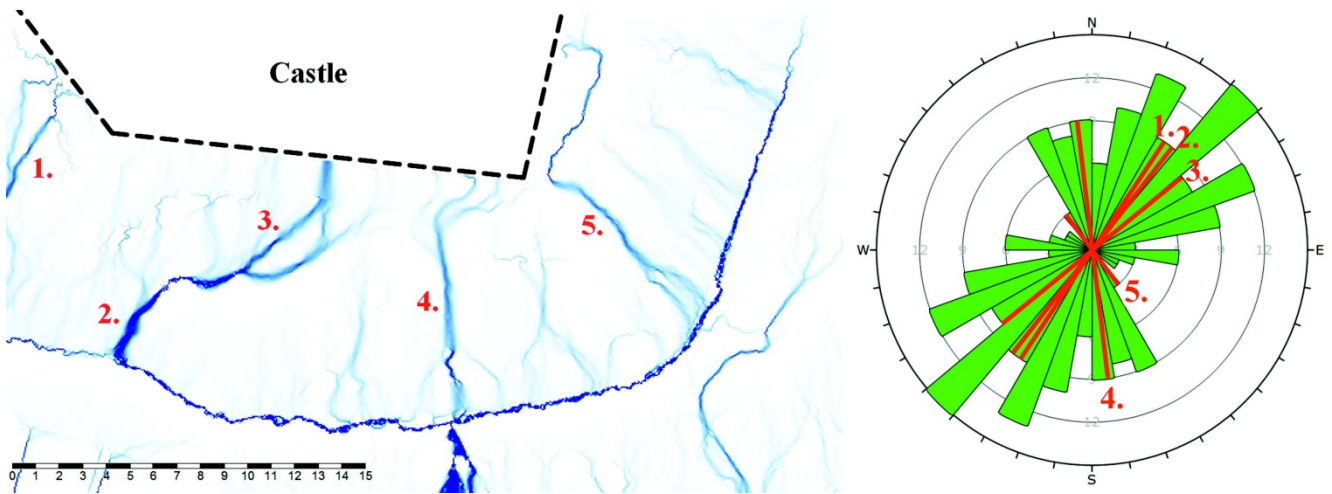
215 Combining and comparing all measured data of discontinuities and joints, using DSM and its derivative of (Fig 6.) and
 216 morphological index (Fig. 9) the joint orientation was outlined. The filed survey validated the obtained structural geological
 217 conditions and six main joint sets (with dip angle/dip directions of: 85/156, 88/312, 79/110, 81/089, 82/064, 61/299) were
 218 identified with prevailing NE – SW direction.

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220



221



222

223 Fig.9. Top view of the cliff (see the location on Fig. 3) obtained with RPAS and the catchment area diagram obtained from DSM
 224 analysis by using Catchment Area module (Costa-Cabral et al. 1994). The latter one was used for joint pattern recognition.
 225 Numbers refer to major joint systems marked on catchment area map and on rose diagram of the field measurements and DSM
 226 data set.

227

228 The laboratory tests of tuffs provided the input data for stability analysis for the two main lithologies: upper and lower unit
 229 of lapilli tuff and middle unit of less welded tuff (Table 3). In the model calculations GSI=50 value was used.

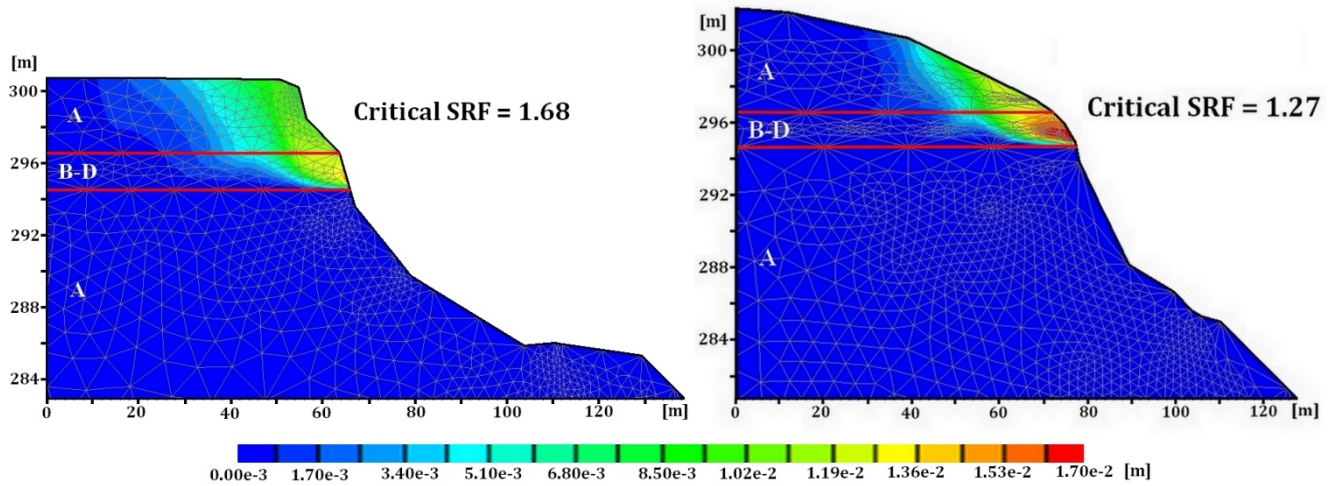
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231 Table 3. Rock mechanical parameters of tuff used in the model: lapilli tuff refers to upper and lower 4 metres, less welded tuff
 232 refers to middle stratigraphic unit

Mechanical property		Upper and Lower unit (marked by A on Fig. 10) (Lapilli tuff)	Middle unit (marked by B-D on Fig. 10) (Less welded tuff)
Bulk density(ρ)	[kg/m ³]	1815	1635
Uniaxial compressive strength(σ_c)	[MPa]	8.02	0.35
Tensile strength (σ_t)	[MPa]	0.83	0.04
Modulus of elasticity(E)	[GPa]	0.97	0.05

233

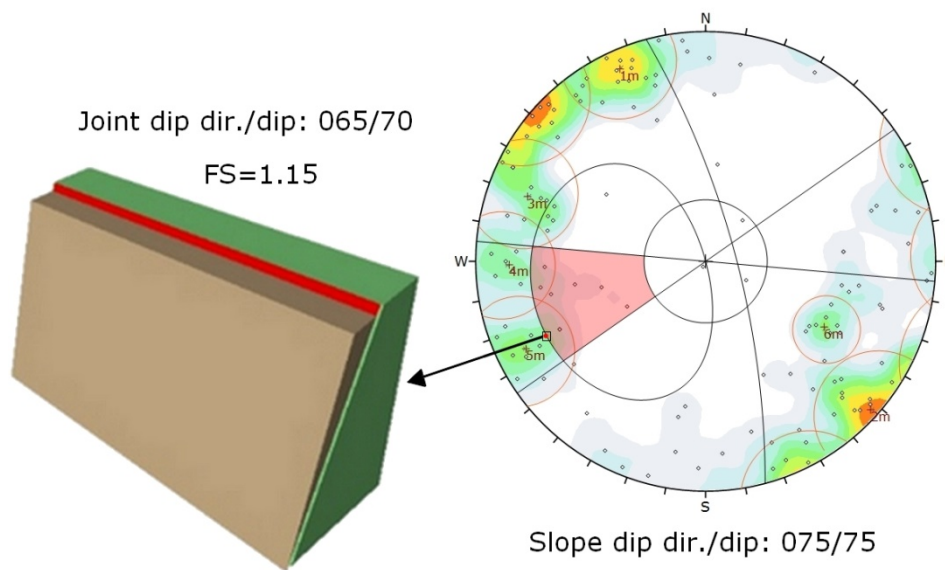
234 The results of RS2 FEM analyses suggest that the global factor of safety is $SRF=1.27-1.71$ in the studied sections (some of
 235 the selected sections are shown on Fig 3). The aim of the analysis is to determine the critical strength reduction factor (SRF)
 236 that can be considered the safety factor of the slope (Fig. 10). The SRF factor is influenced by the weak tuff layer (marked
 237 by B-D on Fig. 8), which has very low shear strength compared to the lapilli tuff. Colours on Fig. 10. represent the total
 238 displacements as a result of the shear strength reduction analysis (Rocscience, 2017). Thus these are not real displacements
 239 of the hillslope. The figure demonstrates only how the failure of the slope would occur with reduced shear strength
 240 parameters. Our failure analyses have proved that the bottom of the slip surface would be in this weaker layer and could lead
 241 to a larger mass movement.



242
 243 **Fig.10. The results of the global stability analysis of the slopes (sections 3 and 4 on Fig. 3), total displacements are marked in blue**
 244 **to red (lithology is indicated by letters A-D, note the weak zone marked by B-D, description of lithologies is given on Fig.8.)**

245
 246 Other failure modes that were studied are planar failure and wedge failure, which are often controlled by joints and
 247 discontinuities. According to data obtained from remote sensing and according to field measurements there was no uniform
 248 spacing of the discontinuities. Stereographic plots with possible failure planes for all slope directions (Fig. 11) indicate that
 249 the most hazardous part of the slope is the one where the plane orientation is 75/75. The calculated factor of safety ($FS=$
 250 1.15) implies high probability of planar failure.

251



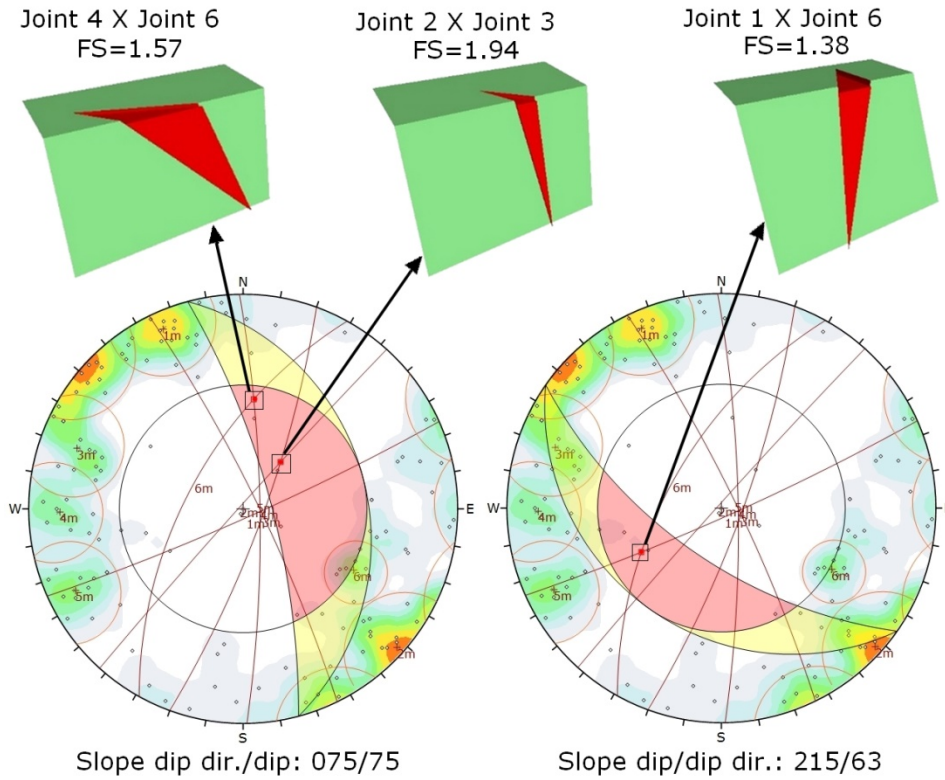
252
 253 **Fig.11. Kinematic analysis of planar failure by Rockplane (main joint sets are marked by red circles 1m-6m)**

254

255 Three possible wedge failure modes were identified as being the most hazardous, according to our calculations by Swedge
256 software (Fig. 12). In these three cases the factor of safety was in the range of 1.38-1.94 representing the hazards of rock
257 falls along wedges delineated by different joint systems.

258

259



260

261 **Fig.12. Examples for the kinematic analysis of wedge failures (main joint sets are marked by red circles 1m-6m)**

262

263 5. Discussion

264 There are three critical sets of input data in the modelling of rocky slopes: i) terrain model and slope geometry, ii) joints
265 system and iii) strength of rock mass.

266 To obtain the first, the slope geometry, RPAS-based surveying technique was used similarly to previous studies (Giordan et
267 al. 2015). In many previous studies RPAS mission was performed following a flight plan (Eisenbeiss 2008, Lindner et al.
268 2016). In our case no flight plan was made prior to mission, and the windy conditions were also not favourable for pre-
269 programmed flight route. The flight was manually controlled and the skilled personnel with a First-Person Viewer tool
270 controlled the image acquisition. The crucial points were the necessary overlaps of images and the matching. The overlaps
271 were ensured by three consequent flights around the study area that provided a dense network of image acquisition locations
272 (Fig. 5). The obtained 1174 images covered the entire study area with appropriate overlaps. The number of images is
273 reasonable, since in previous studies 400 images were taken for a smaller translational rockslide by a GoPro Hero 3 Black
274 camera (Tannant et al. 2017) or app. 400-900 images with a higher resolution camera (18MP) for a landslide area that is
275 approximately five times larger than this one (Lindner et al. 2016). The Pix4Dmapper software (SfM) was used to identify
276 keypoints. Nearly 10,000 keypoints were found on each image (Table 1), which is considered a sufficient amount to provide
277 appropriate matching (Remondino et al. 2014). Camera self-calibration tool and rolling shutter effect correction were the

278 other key features of this software that allowed the processing the images of GoPro Hero 3+ camera. The obliquity of images
279 (Rupnik et al. 2014) and the density of obtained data (Remondino et al. 2014, Rupnik et al. 2015) are crucial in the
280 applicability and accuracy of these images. These were also managed by Pix4Dmapper. The GNNS system and the ground
281 control points (Fig. 6) allowed to georeferencing of the rocky slope. Our results show that the mean 3D accuracy was 4.9 cm.
282 It is comparable with the ground resolution of 1-3.5 cm/pixel of Italian rockslide survey (Tannant et al. 2017), or 3.3-4.1 cm
283 (Neugirg et al. 2016). This resolution was appropriate to create a reliable Digital Surface Model.

284 RPAS-based data was also validated with TLS measurements. The co-use of these remote sensing tools has been previously
285 well-documented for other applications such as soil roughness (Milenkovich et al. 2016) in erosion detection (Neugirg et al.
286 2016) or in cultural heritage (Eisenbeiss & Zhang 2006). The RPAS obtained data validation was performed by comparing
287 the two point clouds obtained by RPAS and TLS. The surfaces were resampled in order to homogenize the spatial resolution.
288 The point densities have been tested in CloudCompare, as a unit sphere of volume of 1 m³ was defined where the points can
289 be counted and then the sphere can be moved along the whole surface. The differences in point clouds are less than 10 cm
290 (Fig 7), which is considered a reasonably good match in terrain modelling (Neugirg et al. 2016). This computation proved
291 that the average point density in both point clouds are practically the same, although the RPAS densities are more
292 homogeneous, while the TLS has denser point clouds close to the scanning stations, as it was expected on the basis of
293 previous works (Naumann et al. 2013)

294 Another aspect causing some differences between the two data sets is that the image based reconstruction is performed by
295 interest operators, very usually SIFT (Scale-invariant Feature Transform) or similar computer vision operators (Lowe 2004).
296 These operators are generally sensitive to intensity jumps, points, or corners, and textural changes in the input images. If the
297 image resolution is not adequate or the object is locally “smooth”, these operators do not return with surface points and the
298 output of the reconstruction has some “filtered” effect. Fortunately, the surface reconstruction quality in RPAS processing
299 resulted minor, ignorable smoothing effects. Comparing the two data sets, it is clearly proven that the geometric resolution of
300 the RPAS-based digital surface model corresponds to the TLS one, offering very similar quality data (Fig.7). It is necessary
301 to note that vegetation can hamper TLS measurements (Prokop & Panholzer 2009, Scaioni et al. 2014, Tannant 2015) and
302 thus limit the comparison of RPAS and TLS obtained data (Milenkovic et al. 2016). In our case most of the study area was
303 bare and if vegetation occurred manual removal was made.

304 The documentation of joint system and discontinuities are crucial for rock wall stability assessment (Tannant et al. 2017).
305 Field survey can only provide reliable data on joint orientation of accessible areas (Margottini et al. 2015); however, the joint
306 system that was found on inaccessible cliffs was not detectable. To overcome this problem, RPAS generated images were
307 used; the frequency of joints was observed based on these images like in previous studies by Assali et al. (2014), Martino &
308 Mazzanti (2014) and Margottini et al. (2015). The required resolution for joint frequency is in the order of 10 cm, rarely 1
309 cm (Tannant 2015, Tannant et al. 2017). The RPAS technique allows for plane surface geometries, however, many joints are
310 not plane surfaces and there are sets in shadows that are difficult to visualize. Thus, RPAS can be used to outline the strike of
311 major joints, but it might cause problems when it comes to the determination of dip and the displacement along the fault
312 planes (e.g. slickensides).

313 Our field tests indicate that the application of Schmidt hammer in rock strength analysis is limited when it comes to the
314 analysis of low strength rocks, such as volcanic tuff (Aydan & Ulusay 2003). As a consequence, laboratory analyses of
315 samples were also required to obtain reliable strength parameters. To measure the strength and to understand the weathering
316 characteristics, samples were taken, representing different stratigraphic positions. Our lab test data (Table 3) clearly indicates
317 that a low strength unit is found in the studied sections (unit marked by ‘B-D’ on Fig 8). Whether the low strength of this
318 zone is related to differential weathering (Török et al. 2007), or it is associated with inherited weakness (micro-fabric), is not
319 clear. This layer is a potential failure zone as it was shown by slope stability calculations. Similar intercalation of pumice-
320 rich layered deposits was modelled by Damiano et al. (2017). They found that a pumice-rich weak zone is prone to rainfall

321 induced landslides. Our results are in good correlations with these findings since the studied rhyolitic volcanic tuff was also
322 proved to be very prone to weathering. A loss in tensile strength of 60% was measured under simulated laboratory conditions
323 (Stück et al. 2008). Weathering processes have been long known to induce landslides and cause slope stability problems in
324 various lithologies and especially pyroclastic rocks (Chigra 2002, Fanti et al. 2013). At the studied rhyolite tuff cliff face, it
325 was shown that joint system is responsible for slope instabilities: planar and wedge failures were found (Fig. 11., Fig. 12).
326 These failure modes are common in hard jointed cliff faces (porphyry Agliardi et al. 2013, mica-schist Tannant et al. 2017,
327 limestone Feng et al. 2014). Our study demonstrates that joint system has an important control on slope stability not only in
328 hard rock lithologies but also in weak tuffs. It is in the line with Fanti et al. 2013 and Margottini 2015, since in Italy and in
329 Giorgia rock walls of volcanic tuffs suffered landslides. The kinematic analysis of tuff rock walls of Tuscany (Fanti et al.
330 2013) also demonstrated that wedge failure and planar failure are the most common failure mechanisms of tuff cliff faces.
331

332 **6. Conclusions**

333 - The manually controlled flights of RPAS provided excellent information on slope geometry of highly dissected and
334 inaccessible slopes.

335 - The necessary overlap between images was ensured by three flights over the small area, by the skilled personnel using
336 First-Person Viewer system with a synchronous image transfer.

337 - The obtained data were managed by Pix4Dmapper (SfM) software allowing the identification of nearly 10.000 keypoints per
338 image.

339 - The TLS based point clouds proved to be good tools to validate the accuracy of images and data sets of manually controlled
340 RPAS. In our study the maximum difference between the two point clouds were less than 10 cm, but mostly around 1 cm.

341 - RPAS collected images and the point cloud based Digital Surface Model and especially Catchment Area method allows the
342 detection of joint system (mainly strikes and partly dips but not slickensides) but field validation and field measurements of
343 accessible joints and faults are recommended to justify joint orientation.

344 - The obtained digital surface model was accurate enough to allow complying cross-sections for rock wall stability
345 calculations.

346 - The lithology and physical parameters of the studied steep cliffs are not uniform and intercalations of weak layers of vitric
347 tuff and volcanoclastic breccia were found.

348 - According to 2D FEM modelling the intercalating low strength layer is the one where potential slip surface can develop
349 causing larger scale mass movements, but at present it has low probability.

350 - Joint system has a crucial role in the stability of the studied rhyolite tuff cliff faces. The highest hazard is related to planar
351 failure along ENE-WSW joints and to wedge failure.

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357

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