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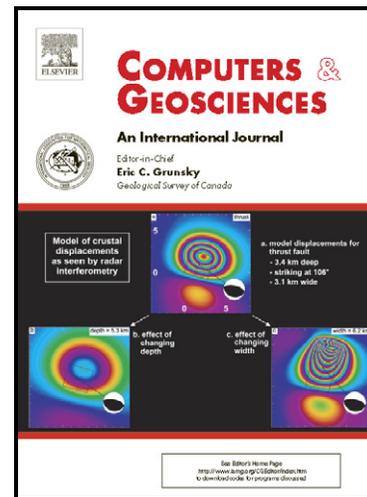
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### **Multiview 3D reconstruction in geosciences**

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**ABSTRACT**

Multiview three-dimensional (3D) reconstruction is a technology that allows the creation of 3D models of a given scenario from a series of overlapping pictures taken by using consumer-grade digital cameras. This type of 3D reconstruction is facilitated by freely available software which does not require expert-level skills. This technology provides a 3D working environment which integrates sample/field data visualization and measurements tools. In this study, we test the potential of this method for 3D reconstruction of decimeter-scale objects of geological interest. We generated 3D models of three different outcrops exposed in a marble quarry and two solids: a volcanic bomb and a stalagmite. Comparison of the models obtained in this study using the presented method with those obtained by using a precise laser scanner shows that multiview 3D

36 reconstruction yields models that present a root mean square error/average linear  
37 dimensions between 0.11 and 0.68%. Thus this technology turns out to be an extremely  
38 promising tool which can be fruitfully applied in geosciences.

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## 42 <h1>1. Introduction

43

44 Multiview 3D reconstruction is the computationally complex process by which a full  
45 3D model of a target scene is derived from a series of overlapping pictures of the target  
46 itself. The method lies at the frontier of computer vision research, and relies also on older  
47 methods used in photogrammetry (Mikhail et al., 2001). The large distribution of high-  
48 resolution ( $\geq 10$  megapixels) consumer-grade cameras and the free availability of open-  
49 source programs implementing structure from motion (SfM) methods make 3D  
50 reconstruction from multiview simple and low cost. SfM is a process used to estimate  
51 both the scene geometry and the camera parameters (Hartley and Zisserman, 2004).  
52 Intrinsic camera parameters are either known a priori (Nister, 2004) or recovered a  
53 posteriori through autocalibration (Triggs, 2000). In a typical SfM procedure, the first  
54 step is the identification of distinctive features (key points) in the input images. Then a  
55 bundle adjustment algorithm allows the reconstruction of the 3D geometry of the scene by  
56 optimizing the 3D location of key points, the location/orientation of the camera, and its  
57 intrinsic parameters (Lourakis and Argyros, 2008; Triggs et al., 2000).

58 Recently, point clouds produced by bundle adjustment methods have been widely  
59 used to create models of architectural and bare earth surfaces with high accuracy (de  
60 Matías et al., 2009; Dowling et al., 2009; Grzeszczuk et al., 2009). Similar high-  
61 resolution 3D photorealistic models of geological outcrops constitute virtual outcrops  
62 which are ideal for visualization and quantification of 3D structural or sedimentary  
63 features, maximizing the benefit of field excursions (Bellian et al., 2005; McCaffrey et  
64 al., 2005; Pringle et al., 2006; Buckley et al., 2008). Photogrammetric surveys and  
65 computer vision techniques have been also used by James et al. (2007) to characterize  
66 morphological modifications of an advancing lava flow.

67 In this study we evaluate the performance of the multiview 3D reconstruction method  
68 in geosciences. We create 3D models of three outcrops exposed on a marble quarry and  
69 two solid samples of geological interest: a volcanic bomb and a stalagmite. Our examples

70 have typical linear dimension up to 1 m. For this purpose we defined a sequence of  
71 automatic steps which only uses freely available software and does not require any prior  
72 information on camera position, orientation, or internal camera parameters. The accuracy  
73 of the obtained models is assessed by comparison with models obtained by using a laser  
74 scanning technology.

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76

## 77 **<h1>2. Methods**

78

### 79 *<h2>2.1. Multiview 3D reconstruction*

80

81 Multiview 3D reconstruction creates a 3D model starting from a series of overlapping  
82 photos imaging a given scene. This is achieved by running a series of algorithms which  
83 work automatically without a priori specification of parameters for the input pictures. The  
84 procedure applied in this work comprises the following steps: (i) the scale invariant  
85 feature transform (SIFT) algorithm (Lowe, 2004) is used for key-point extraction; (ii) the  
86 open-source SfM software package Bundler (Snavely et al., 2006, 2007) generates a  
87 sparse 3D point cloud with internally consistent 3D geometry; (iii) the open-source  
88 PMVS2 (Patch-based Multiview Stereo software – version 2) software takes the output of  
89 the Bundler software as input to reconstruct the model of the imaged scene in the form of  
90 a denser point cloud (Furukawa and Ponce, 2007, 2009); (iv) additional software is used  
91 for visualization and postprocessing. In the following sections, more details are provided  
92 for each step.

93

#### 94 *<h3>2.1.1. Recommendations for photoacquisition*

95 The sequence of pictures, which constitute the starting input, must be taken from  
96 several viewpoints which vary significantly from one another. As an example, many  
97 pictures from the same viewpoint are useless, while pictures taken at each step by moving  
98 around the scene of interest are ideal. The sequence of pictures must be acquired while  
99 the target scene/object is fixed in the same position under a good lighting, and moving  
100 shadows and/or camera flash should be avoided as much as possible. In addition, the  
101 color texture (nonhomogeneity) of the object/scene of interest is important, because the  
102 procedure works on color changes. The theoretical minimum number of input photos is 3,  
103 but a minimum of 4 to 6 pictures is recommended to obtain reliable models, and the

104 model accuracy increases if a much higher number of “good” pictures is used (from tens  
105 to hundreds of pictures). An example of a good sequence of viewpoints (one picture from  
106 each viewpoint) is given in Fig. 1.

107

### 108 *<h3>2.1.2. Feature extraction*

109 In the first step of the global procedure, all the pictures are processed in loop by a  
110 pattern recognition algorithm and matched to each other to find corresponding features in  
111 different images. In this way a series of key points is obtained. This process is carried out  
112 by using the scale invariant feature transform (SIFT) algorithm (Lowe, 2004). A demo  
113 version of SIFT is available at <http://www.cs.ubc.ca/~lowe/keypoints/>. This demo (at  
114 present) works only on small images (a few megapixels); hence we down-sampled input  
115 images to fulfill this constraint. To simplify the processing, the input pictures are  
116 converted into gray scale images before running the SIFT algorithm. Regions of interest  
117 are those marked by sharp gradients in gray values. Typically, SIFT will detect up to tens  
118 of thousands of such features in a resampled image.

119

### 120 *<H3>2.1.3. Structure from motion processing*

121 Once corresponding key points have been identified across a series of images, the  
122 change in key-point position in different images is considered in the SfM process to  
123 clump the position of such points in a 3D reference system. This complex process takes  
124 also into account the focal length and sensor width of the camera used to take the image  
125 (the camera type is tagged in the header of the picture file). The output provides camera  
126 parameters and position for each considered input image by using a numeric optimization  
127 technique called “bundle adjustment.” In this work we use Bundler software  
128 (<http://phototour.cs.washington.edu/bundler>) which is an open-source SfM software  
129 (Snavely et al., 2006) that iteratively considers an increasing number of input pictures,  
130 providing an increasingly optimized output as the process goes on. If an input image is  
131 “not good” (e.g., it is blurred) Bundler automatically discards it. Bundler outputs also a  
132 sparse cloud of 3D points representing the imaged scene.

133

### 134 *<H3>2.1.4. Dense 3D point cloud reconstruction*

135 The output obtained from Bundler is then processed by the Patch-based Multi-View  
136 Stereo – version 2 package (PMVS2, Furukawa and Ponce, 2007; 2009). An open-source  
137 implementation of PMVS2 is available at <http://grail.cs.washington.edu/software/pmvs/>.

138 This further processing produces a much denser 3D point cloud which provides a very  
139 detailed and realistic model of the imaged scene. One of PMVS2 advantages is that it  
140 preserves only rigid structures (e.g., pedestrians walking in front of a monument will not  
141 be seen in the final result). PMVS2 is also robust against differences in image colors due  
142 to exposure settings, white balance, or lighting conditions. Various parameters and flags  
143 can be specified in the PMVS2 option file including the subsampling rate of images  
144 before the processing; a tentative density of reconstruction; the minimum number of  
145 images in which a point must be visible to be reconstructed; the minimum photometric  
146 consistency measure necessary to keep a point in the reconstruction (for details, see  
147 <http://grail.cs.washington.edu/software/pmvs/documentation.html>).

148

149 <H3>2.1.5. Visualization, surface reconstruction, and postprocessing

150 In the case of simple and substantially flat geometries, as for most outcrops,  
151 postprocessing can be done in a GIS environment by treating the obtained 3D models as  
152 digital elevation models (DEMs) with x and y coordinates assigned along the plane fitting  
153 the sampled surface and the elevation set orthogonally to this plane.

154 In the case of more complex 3D geometries, such as the two solid samples, a series of  
155 freely available tools have been used for the postprocessing, the rendering, and the error  
156 assessment of the obtained 3D models. 3D point clouds have been managed using the  
157 Scanalyze software, developed by the Stanford Computer Graphics Laboratory, freely  
158 available at <http://graphics.stanford.edu/software/scanalyze/>. Scanalyze is a computer  
159 graphics program for viewing, editing, and merging range images to produce denser  
160 polygon meshes (Besl and McKay, 1992; Levoy et al., 2000). The open-source MeshLab  
161 software has been used to connect the points cloud generated by PMVS2 in a network of  
162 triangles which approximates the continuous surface of the imaged scene. MeshLab  
163 allows the editing of unstructured 3D triangular meshes. This freely available software  
164 has been developed by the Visual Computing Lab of ISTI-CNR in Pisa, Italy  
165 (<http://meshlab.sourceforge.net/>). Finally, to compare the difference between pairs of  
166 complex surfaces we have used Metro, a tool designated to evaluate the difference  
167 between two triangular meshes (Cignoni et al., 1998). The mean distance  $E_m$  of a surface  
168  $S_1$  from a surface  $S_2$  is defined as the surface integral of the distance divided by the area  
169 of surface  $S_1$ ,

170

$$E_m(S_1, S_2) = \frac{1}{|S_1|} \int_{S_1} e(p, S_2) ds, \quad (1)$$

172 where  $e(p, S_2)$  is the distance between a point  $p$  (belonging to  $S_1$ ) and the surface  $S_2$ .  
 173 Indeed Metro compares two triangular meshes  $S_1$  and  $S_2$  numerically (see Cignoni et al.,  
 174 1998, for further details).

175

176 <h2>2.2. Laser scanning reconstruction

177

178 3D control models of the selected test surfaces have been obtained by using Konica  
 179 Minolta VI-910 laser scanning, a noncontact 3D digitizer ([www.konicaminolta-3d.com](http://www.konicaminolta-3d.com)).  
 180 The target surface is scanned by a laser beam (wavelength = 690 nm) emitted from the  
 181 VI-910's source and the signal reflected back by the target is captured by the VI-910's  
 182 CCD receiver. Coordinates (x, y, and z) of imaged objects are reconstructed through  
 183 triangulation. This device stores a mesh of 640×480 3D points at each acquisition. The  
 184 VI-910 is provided with three interchangeable lenses to fit a variety of scanning settings.  
 185 A single acquisition captures an area between ~10 cm<sup>2</sup> (TELE lens) and ~0.8 m<sup>2</sup> (WIDE  
 186 lens). The instrument maximum accuracy is achieved using the TELE lens: 0.22 mm in x,  
 187 0.16 mm in y, and 0.1 mm in z, the z axis being the optical axis of the laser scanner. For  
 188 this work we used only the WIDE lens which has accuracies of 1.4 mm along x, 1.04 mm  
 189 along y, and 0.4 mm along z. 3D models are created using the Konica Minolta Polygon  
 190 Editing Tool by data alignment, merging, and triangulation. No successive filling or  
 191 smoothing was performed.

192

193

194 <h1>3. Test cases

195

196 <h2>3.1. Outcrops

197 We selected three sites (S1, S2, and S3) exposed on a subvertical fresh outcrop in a  
 198 marble quarry located on the South flank of Mt. Castellare, near San Giuliano Terme  
 199 (Pisa, Italy; Figs. 2a, b, and c). From a geological perspective the area belongs to the  
 200 Monti Pisani Unit, one of the main metamorphic outcrops of the Northern Apennine, that  
 201 has been subjected to two main episodes of deformation, a first compressive ductile phase  
 202 between the late Oligocene and the early Miocene followed by an extensional phase  
 203 during the Tortonian (Carosi et al., 2004, and references therein). This poly-phase

204 deformation history results in a complex pattern of fractures clearly visible on the quarry  
205 surface. The three sites show various chromatic and textural characteristics representing  
206 different geological aspects, despite their collocation a few tens of meters apart.

207 At site S1 the liassic marble "Calcare ceroide" crops out (Rau and Tongiorgi, 1974). It  
208 is a low-grade metamorphic white, gray or whitish-yellowish marble with thin layers of  
209 muscovite (Figs. 2a and 3). The quarry cuts small cave passages unearthing physical and  
210 chemical cave deposits. At site S2 different speleothems are present (Figs. 2b and 4): (i) a  
211 thin flowstone originated as a calcite deposit from a uniform water flow and accreted  
212 roughly parallel to the surface (almost vertical in this case); (ii) a small stalactite (i.e., a  
213 subvertical concretion growing from top to bottom as a result of carbonate deposition  
214 from water drops); and (iii) cave popcorn concretions (i.e., globular calcite deposits  
215 developed in a low evaporation environment). At site S3, a small sedimentary breccia  
216 section crops out. It is an unsorted debris, grain supported, probably derived from a  
217 colluvium tongue, transported in depth by gravity through a fracture (Figs. 2c and 5).

218 At each site we collected a large series of pictures suitable for the multiview 3D  
219 reconstruction procedure, by using a Canon EOS 450D digital camera. The same scenes  
220 have been imaged by using the Konica Minolta VI-910 laser scanner mounted with the  
221 TELE lens. 3D models of sites have been built from both acquisition systems.

222 The acquired areas are approximately rectangular and cover extents between  $\sim 0.15$   
223 and  $\sim 0.3 \text{ m}^2$  (average linear dimensions from  $\sim 40$  to  $\sim 55$  cm, see Figs. 3, 4, and 5, Table  
224 1). To explore the effectiveness of the multiview method with respect to the series of  
225 input pictures, the models of the three outcrops have been derived by processing a  
226 different number of pictures. The model for site S1 was obtained by processing four  
227 photos, producing a final cloud of  $\sim 55,000$  points; the model for site S2 was obtained by  
228 processing 40 photos, resulting in a final cloud of  $\sim 200,000$  points; and the model for site  
229 S3 by processing 35 photos, and a final cloud of  $\sim 450,000$  points.

230 The point cloud density is clearly related to the number of input photos but also to  
231 their quality and to the acquisition geometry. In fact models of S2 and S3 have been  
232 reconstructed starting from a similar number of pictures (40 vs 35) but the average  
233 number of points per photo in the final point clouds is rather different ( $\sim 200,000$  vs  
234  $\sim 450,000$ ; Table 1).

235

236 *<H2>3.2. 3D modeling of solids*

237 We analyzed two solids different in shape, color, mineral composition, and geological  
238 meaning: a stalagmite and a volcanic bomb (Figs. 2e and d, respectively). A stalagmite is  
239 a speleothem growing from the floor of a cave caused by the dripping of water rich in  
240 calcium bicarbonate. A volcanic bomb is a lava projectile which by definition is larger  
241 than 65 mm in diameter and ejected by a volcano during an eruption.

242 The stalagmite modeled in this work was taken from the Buca di Cavorso (Jenne,  
243 Roma, Italy). It has the typical tapered shape (Figs. 1 and 6), a height of ~27 cm, and  
244 basal diameter of ~10 cm. A 3D model was reconstructed using 30 photos obtaining a  
245 cloud of ~185,000 points. Comparison with the 3D model obtained using the laser  
246 scanning is shown in Fig. 6 and tabulated in Table 2.

247 The volcanic bomb considered here was ejected during the 2001 eruption at Mt. Etna  
248 (Italy) from the South-East summit crater. This bomb has the typical almond shape (Fig.  
249 7) with the maximum and minimum dimensions of ~15 and ~9 cm, respectively. By using  
250 67 input photos we derived a cloud of ~136,000 points (Table 2).

251

252

#### 253 <H1>4. Discussion

254

255 The outcrops have a simple, almost planar surface and can be reconstructed by using a  
256 small number of photos. On the contrary, the much more complex reconstruction of solids  
257 requires tens of photos. For almost flat surfaces (outcrops S1 to S3) the effective number  
258 of points per photo in the points cloud is high (5000–15,000 points/image for a 1024×638  
259 pixel image; see Table 1); for solids, the number of points per used photo drops  
260 significantly (2000–6000 points/image) despite the simple geometry of the considered  
261 samples (Table 2).

262 For the error assessment, we considered as “ground” truth the 3D models obtained by  
263 using the Konica Minolta VI-910 laser scanner, owing to the low nominal error. The  
264 multiview model of S1 has been derived by using only 4 photos; nevertheless it shows a  
265 low root mean square error (RMSE), though significantly higher than the ones calculated  
266 for the models of S2 and S3. The RMSEs percentage (i.e., RMSE/average linear  
267 dimensions) is 0.68% in the model generated from four photos and 0.11% in the model  
268 generated from 35 photos, which turns out to be the most accurate. The higher error  
269 obtained in the S1 site is easily explained: S1 presents quasi-planar surfaces broken by  
270 big discontinuities and four photos are not able to reconstruct such big discontinuities

271 (e.g., the red area in Fig 3c). Thus a percentage RMSE of 0.68% must be considered a  
272 conservative upper limit rather than the rule since it refers to the worst possible  
273 combination of acquisition geometry and surface characteristics.

274 For the three outcrops, the maps of the depth differences between the models obtained  
275 with multiview 3D and the control models obtained using the laser scanning (Figs. 3, 4,  
276 and 5) show a clear pattern with positive values at the edge of the scenes and negative  
277 ones at the center. This evidence suggests the existence of a systematic error in our  
278 multiview 3D reconstruction.

279 We used the outcrop models (Figs. 3, 4, and 5) to quantify textural differences among  
280 the three sites. We calculated two parameters: (i) the roughness as the root mean square  
281 heights along the viewing direction, and (ii) the detrended roughness, calculated as above  
282 after the subtraction of the best fitting plane from the model. For the purpose of roughness  
283 calculations, the photo-derived triangulated surfaces are “georeferenced” with the  
284 corresponding laser-derived surfaces and then all the pairs of surfaces are converted into  
285 grids. Roughness calculations are performed on the gridded surfaces. Results show that  
286 the detrended roughness of S1 is higher than that of S2 and S3 (16.73 vs 12.56 and 9.79  
287 mm, respectively; Table 1). Percentage errors in detrended roughness, derived by  
288 comparing photo-derived and laser-derived 3D models, are in the range 0.3–2%.

289 For the two solids, we used the software Metro to calculate the errors of the photo-  
290 derived models with respect to the laser-derived models. The stalagmite model has an  
291 overall RMSE of ~0.80 mm, corresponding to a percentage RMSE/average sample linear  
292 dimension of 0.22%. The volcanic bomb model has an RMSE of ~0.33 mm,  
293 corresponding to a percentage RMSE/average sample linear dimension of 0.16% (Table  
294 2). Table 2 shows that RMS distance between the laser-derived and the photo-derived  
295 models can change significantly as the reference solid changes (i.e., the solid from which  
296 the distance is calculated according to Eq. (1)). This is due to missing portion in one of  
297 the models, for example, at the base of bomb in the photo-derived model (Fig. 7). The  
298 photo- and laser-derived models of the stalagmite are more consistent (Table 2 and Fig.  
299 6).

300 Fig. 8 shows the error distributions in all the test cases. S1, S2, and S3 show an  
301 asymmetric distribution which is due to the above-described systematic error (see Figs.  
302 3c, 4c, and 5c). Despite the apparent greater error spreading of S2 and S3, S1 has the  
303 higher RMSE, owing to the biased reconstruction of the discontinuity which cuts almost  
304 horizontally across the sampled surface in the photo-derived model (Fig. 3). For the bomb

305 and the stalagmite we plotted the discrepancies (always positive) between photo- and  
306 laser-derived models.

307 To explore the sensitivity of the method with respect to the PMVS2 settings, we  
308 iteratively rederived all our models introducing small changes in the PMVS2 option file.  
309 We found that these small changes result in negligible variations in points cloud density  
310 and model accuracy.

311

312

## 313 <h1>5. Conclusions

314

315 We assessed the performances of a multiview 3D reconstruction method for  
316 generating full 3D models of small outcrops (areas between  $\sim 0.15$  and  $\sim 0.3$  m<sup>2</sup>) and  
317 decimeter-scale objects of geological interest. The complete processing is carried out by  
318 using only freely available software.

319 Comparisons with reference models acquired by using a laser scanner show that this  
320 method warrants percentage RMSE (RMSE/average sample linear dimension) which can  
321 attain  $\sim 0.1\%$ . Obtained results demonstrate that the multiview 3D reconstruction  
322 technique can be effectively used to substitute much more expensive and cumbersome  
323 technologies (e.g., laser scanners or terrestrial LIDAR) in cases similar to the ones  
324 presented here. The main advantages of multiview techniques are:

- 325 • Simplicity: viable input images can be acquired without any specific competence  
326 and the final 3D reconstruction is straightforward.
- 327 • Flexibility: a multiview survey does not involve logistical efforts because it  
328 requires only the use of a digital camera (easily to bring everywhere).
- 329 • Low cost: multiview 3D reconstruction involves a consumer-grade camera, freely  
330 available software, and the survey does not require additional costs.
- 331 • Scale free: multiview methods are, in theory, not constrained in scale, as long as  
332 the acquired series of pictures fit the required specifications.
- 333 • Acquisition frequency: an acquisition can be as fast as a click: setting up several  
334 cameras in different locations, full 3D acquisition can be done at very short time  
335 steps. This can be very useful, for example, to support laboratory experiments.

336 On the other hand, more expensive techniques can reach higher resolutions and  
337 accuracies and/or work at much longer ranges. Also, lighting conditions affect the final  
338 result, while active acquisition systems do not have similar problems.

339 As a whole, photo-derived 3D reconstructions turn out to be easy, fast, reliable, and  
340 nonexpensive for 3D modeling of scenarios of geological interest. Possible future  
341 applications could include the determination of morphological changes of rapidly  
342 evolving systems (e.g., steep, unstable slopes or riverbeds) and the monitoring of in-  
343 laboratory analog experiments.

344

345

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347

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- 421

422

423 **Figure captions**

424

425 **Fig. 1.** Camera positions and orientations used in the acquisition of the stalagmite. (a) Top  
426 view; (b) lateral view.

427

428 **Fig. 2.** Surfaces used as test cases for the generation of 3D models: (a,b,c) outcrops  
429 exposed on a subvertical fresh wall in a marble quarry located on the South flank of Mt.  
430 Castellare, near San Giuliano Terme (Pisa, Italy; surfaces S1, S2, and S3 of Figs. 3, 4, and  
431 5, respectively, are outlined by red dashed lines); (d) stalagmite from the Buca di Cavorso  
432 (Jenne, Roma, Italy; Fig. 6); (e) volcanic bomb ejected during the 2001 eruption at Mt.  
433 Etna (Italy; Fig. 7).

434

435 **Fig. 3.** Digital model of San Giuliano marble outcrop (site S1). (a) 3D point cloud from  
436 the multiview reconstruction, displayed by RGB color information; (b) slope model of  
437 laser-derived data; (c) difference map between multiview reconstruction and laser-derived  
438 model; (d) slope model of the multiview reconstruction.

439

440 **Fig. 4.** Digital model of calcareous concretion outcrop (site S2). (a) 3D point cloud from  
441 the multiview reconstruction, displayed by RGB color information; (b) slope model of  
442 laser-derived data; (c) difference map between multiview reconstruction and laser-derived  
443 model; (d) slope model of the multiview reconstruction.

444

445 **Fig. 5.** Digital model of small breccias outcrop (site S3). (a) 3D point cloud from the  
446 multiview reconstruction, displayed by RGB color information; (b) slope model of laser-  
447 derived data; (c) difference map between multiview reconstruction and laser-derived  
448 model; (d) slope model of the multiview reconstruction.

449

450 **Fig. 6.** 3D model of a stalagmite: (a) model derived from the multiview reconstruction,  
451 displayed by RGB color information; (b) shaded image of laser-derived model; (c) 3D  
452 difference map between multiview-derived and laser-derived surfaces.

453

454 **Fig. 7.** 3D model of a volcanic bomb: (a,d) model derived from the multiview  
455 reconstruction, displayed by RGB color information; (b,e) shaded image of laser-derived  
456 model; (c,f) 3D difference map between multiview-derived and laser-derived surfaces.

457

458 **Fig. 8.** Error distributions of multiview-derived models of the outcrops (S1, S2, and S3),  
459 the stalagmite and the volcanic bomb considered in this work. Errors are evaluated as  
460 differences with laser-derived models. For the stalagmite and the volcanic bomb, errors  
461 are evaluated as distances between the multiview-derived and the laser-derived surfaces  
462 (Eq. (1)).

463

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466

467 **Table 1**

468 Characteristics of sampled outcrops, laser-derived and multiview-reconstructed models.

469

Parameter		S1	S2	S3
Outcrop extension	Area (m <sup>2</sup> )	0.306	0.169	0.148
	X extent (mm)	642	471	471
	Y extent (mm)	476	359	315
	Average XY scale <sup>a</sup> (mm)	553	411	385
Laser model	N. pts.	269390	226172	189744
	Average mesh step (mm)	1.07	0.86	0.88
	Roughness (mm)	25.90	31.34	16.92
	Detrended roughness (mm)	16.73	12.56	9.79
Photo model	N. photo	4	40	35
	N. pts.	55265	205252	450075
	Average N. pts./N. photo	13816	5131	12859
	Average mesh step (mm)	2.35	0.91	0.57
	Roughness (mm)	25.36	28.89	16.90
	Detrended roughness (mm)	16.68	12.24	9.67
	RMSE <sup>b</sup> (mm)	3.76	1.09	0.41
Percentage error <sup>c</sup> (%)	0.68	0.27	0.11	

470 <sup>a</sup> Calculated as the square root of area.

471 <sup>b</sup> Root mean square error between the laser-derived 3D model and the multiview  
472 3D reconstruction.

473 <sup>c</sup> Calculated as the ratio between the RMSE and the average XY scale.

474

475 **Table 2**

476 Characteristics of laser-derived and multiview-derived models and distance between the  
477 two surfaces.

478

Parameter	Stalagmite		Volcanic bomb	
	Laser	Photo	Laser	Photo
N. vertices	137848	185628	82413	136519
N. faces	269677	320887	156414	236039
Area (mm <sup>2</sup> )	84969	83392	39683	31708
Bounding box diag. D (mm)	368	423	214	204
Max distance (mm)	12.8	14.2	17.6	4.5
Mean distance (mm)	0.54	0.55	0.75	0.23
RMS distance E (mm)	0.81	0.92	2.14	0.33
E/D (%)	0.22	0.22	1.00	0.16

479

480

481

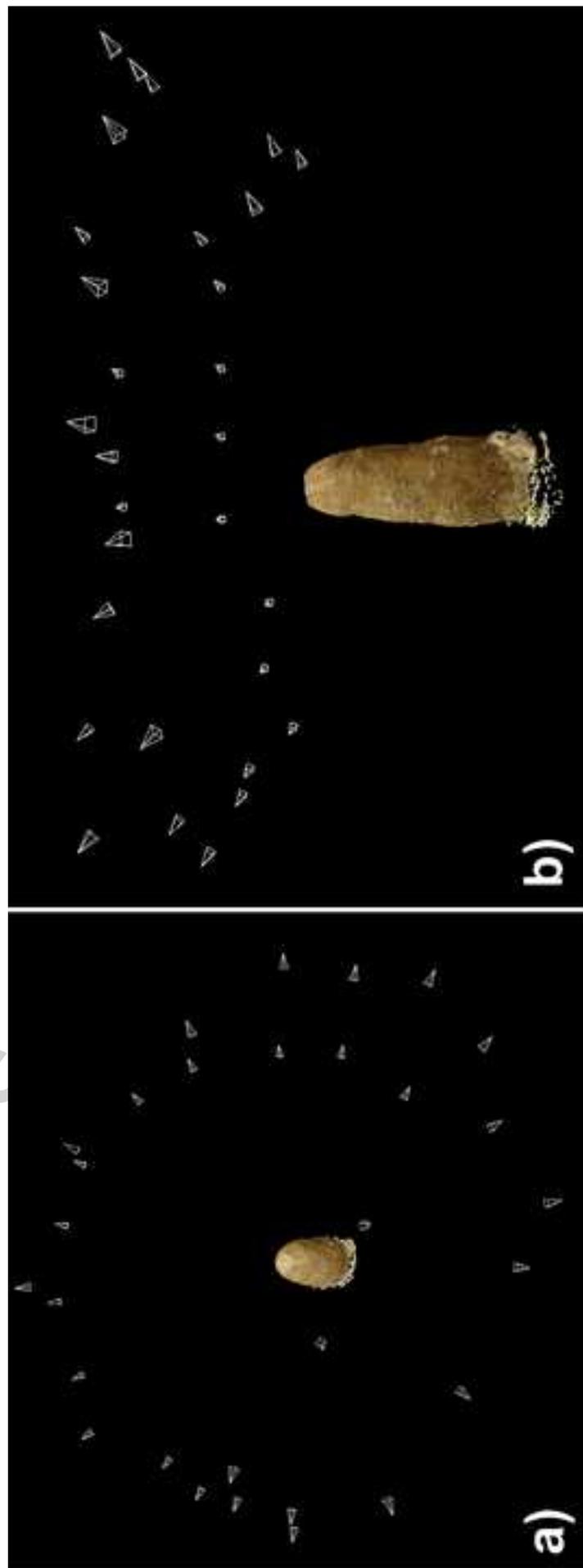
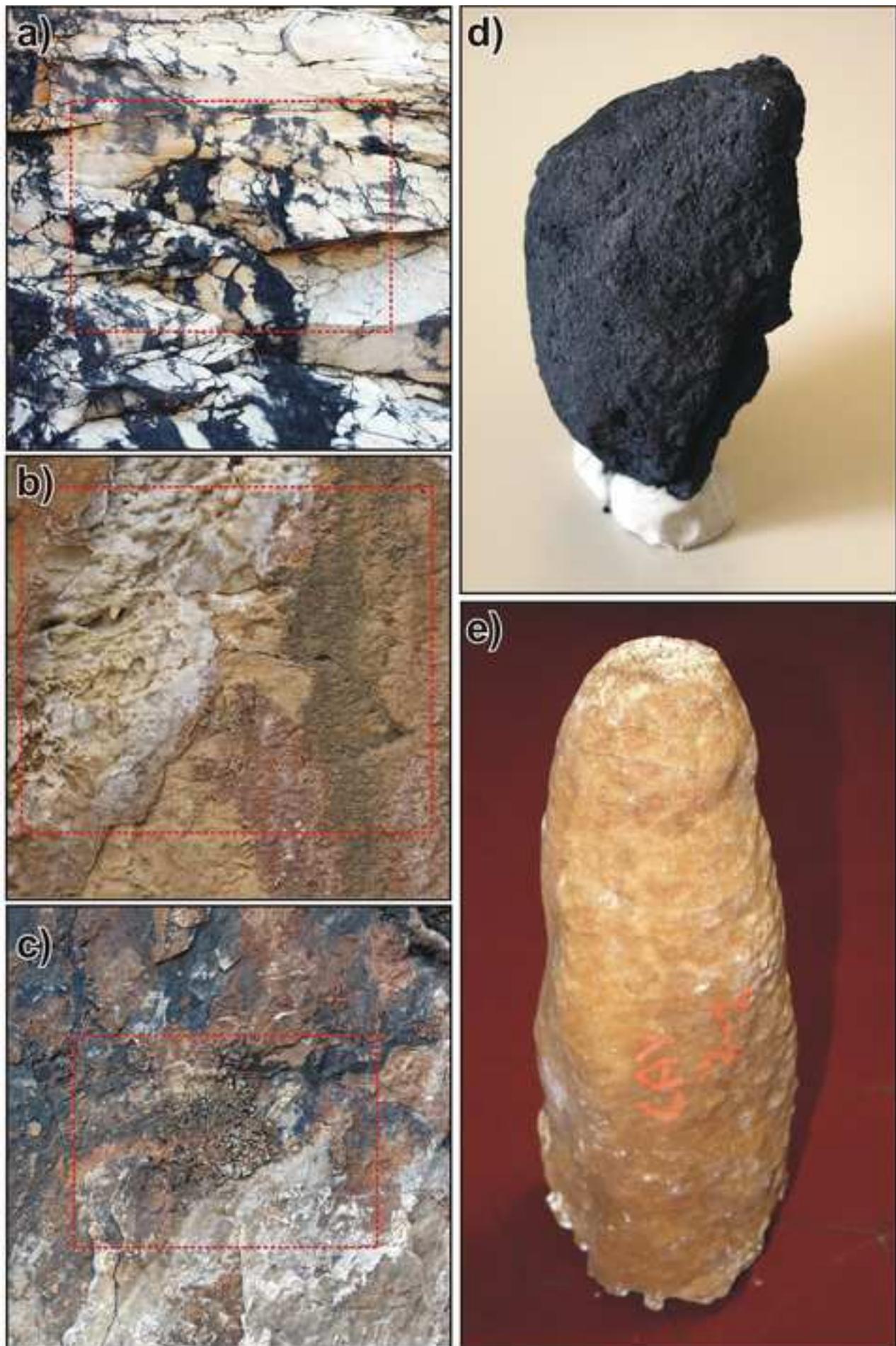


Figure 1



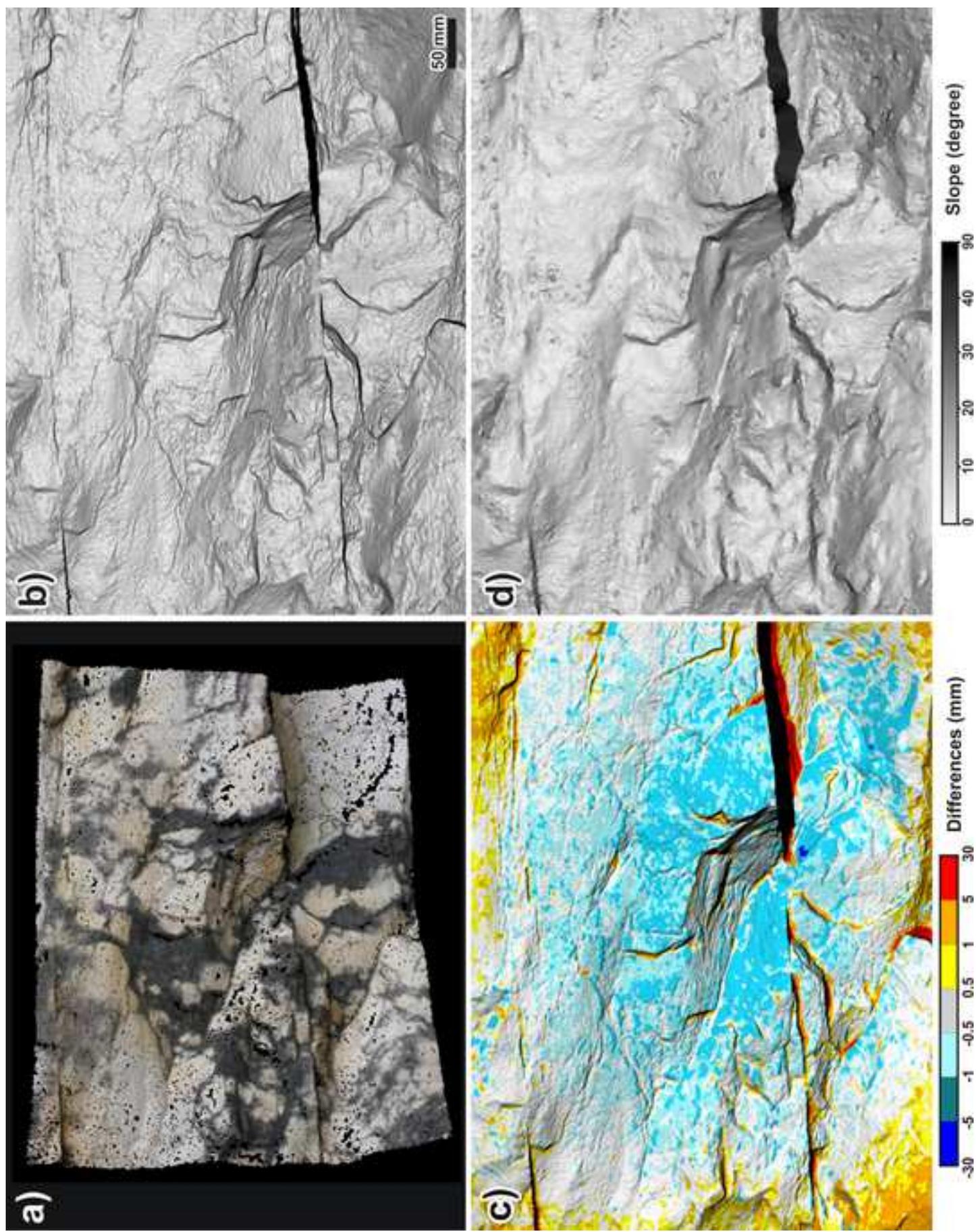


Figure 3

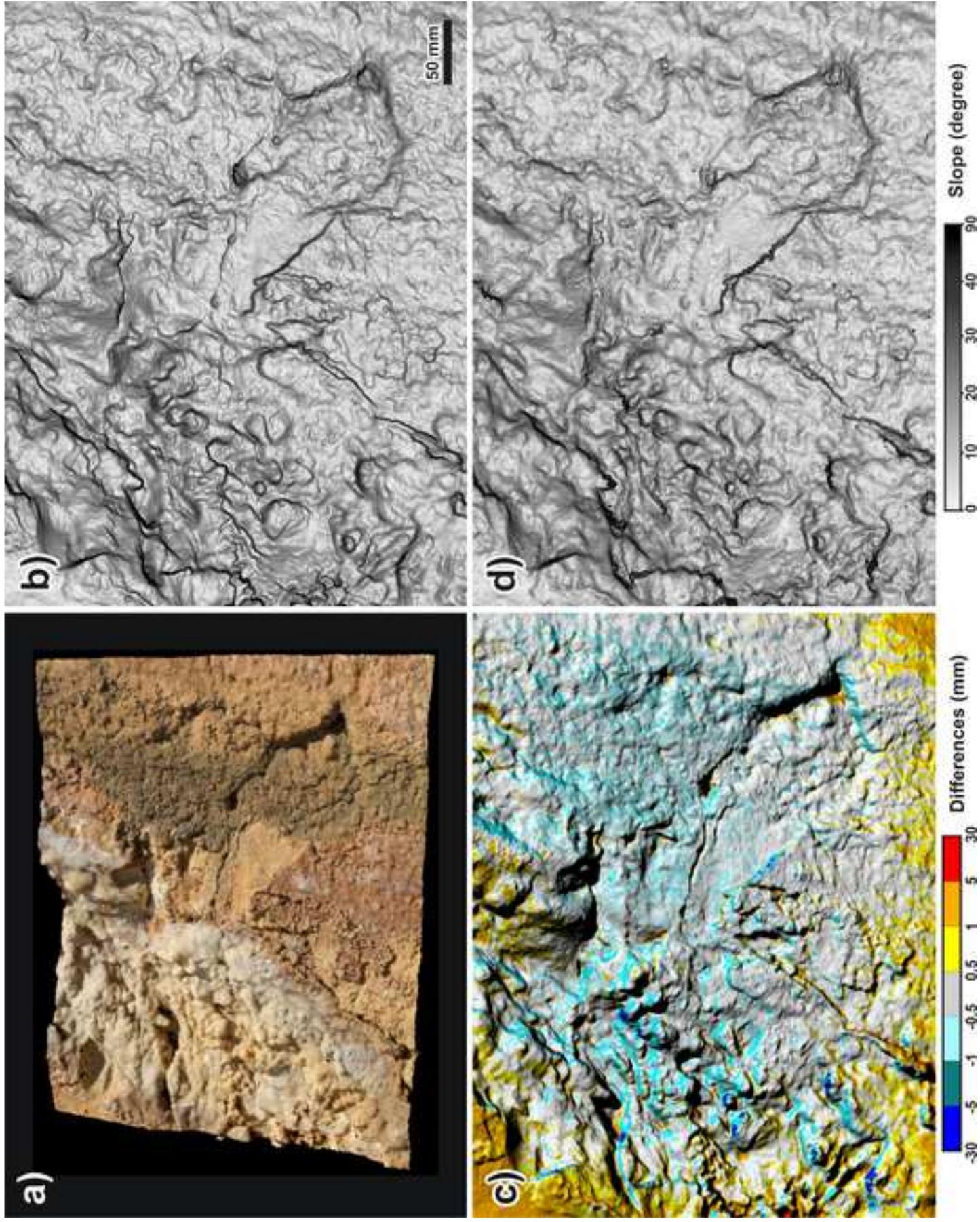


Figure 4

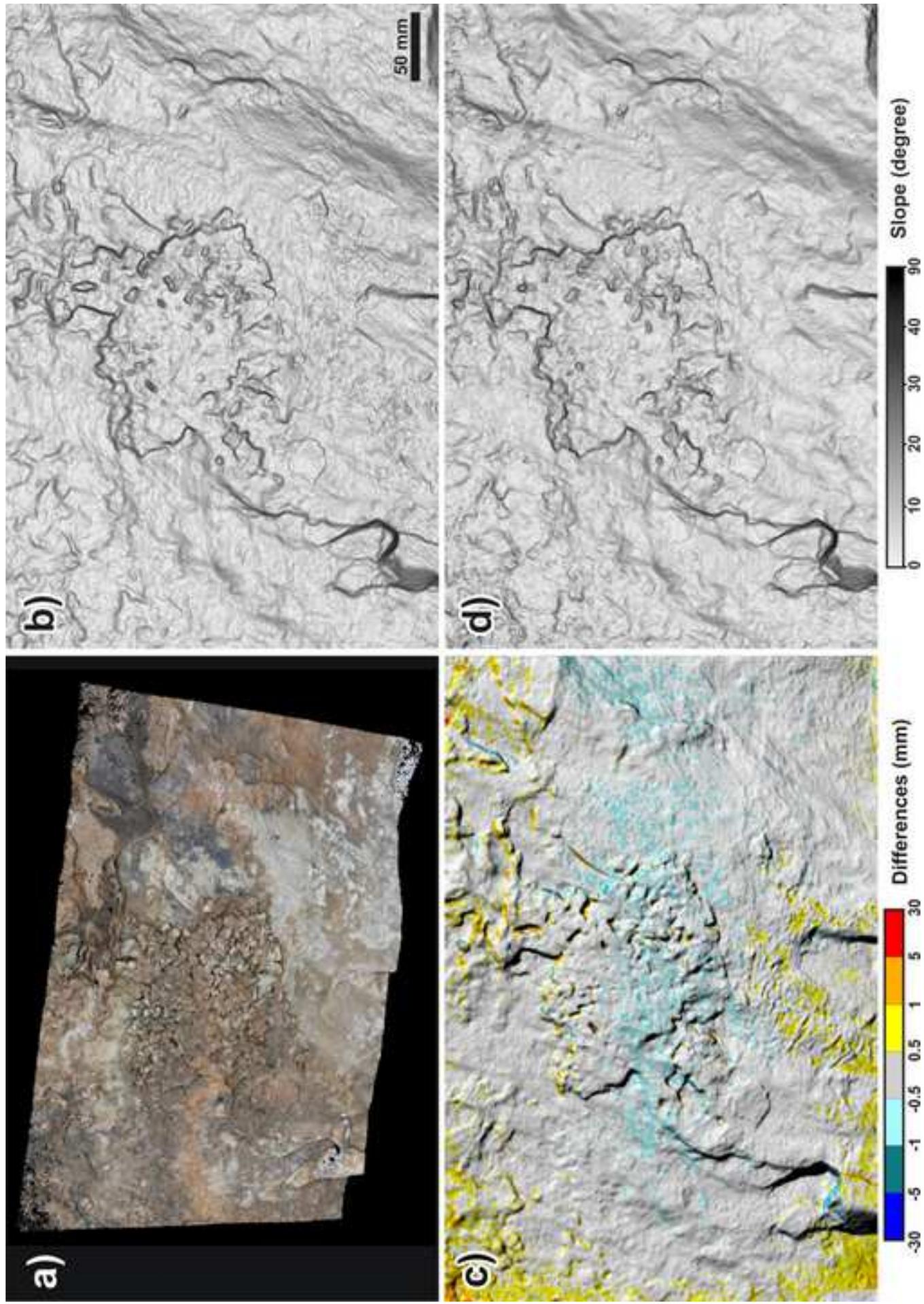


Figure 5

