Monitoring surface deformation with spaceborne radar
 interferometry in landslide complexes: insights from the

**Brienz/Brinzauls slope instability, Swiss Alps** 

4 Andrea Manconi<sup>1,2,3\*</sup>, Nina Jones<sup>1,2</sup>, Simon Loew<sup>1</sup>, Tazio Strozzi<sup>2</sup>, Rafael Caduff<sup>2</sup>, Urs

# 5 Wegmueller<sup>2</sup>

- 6 <sup>1</sup> ETH Zurich, Dept. of Earth Sciences, Engineering Geology, Zurich, Switzerland
- 7 <sup>2</sup> GAMMA Remote Sensing AG, Gümligen, Switzerland
- 8 <sup>3</sup> now at CERC, WSL Institute for Snow and Avalanche Research SLF, Davos, Switzerland
- 9 \*Corresponding author: andrea.manconi@slf.ch

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# 17 Abstract

18 We performed an extensive analysis of C-Band SAR datasets provided by the European Space Agency (ESA) satellites ERS-1/2, Envisat ASAR, and Sentinel-1 in the period 1992-19 2020 aiming at reconstructing the multi-decadal spatial and temporal evolution of the surface 20 displacements at the Brienz/Brinzauls landslide complex, located in canton Graubünden 21 22 (Switzerland). To this end, we analyzed about 1'000 SAR images by applying differential interferometry (InSAR), multitemporal stacking, and Persistent Scatterer Interferometry (PSI) 23 approaches. Moreover, we jointly considered Digital Image Correlation (DIC) on high resolution 24 25 multi-temporal Digital Terrain Models (DTM) generated form airborne surveys and InSAR 26 results to compute 3-D surface deformation fields. The extensive network of GNSS stations across the Brienz landslide complex allowed us to extensively validate the deformation results 27 28 obtained in our remote sensing analyses. Here, we illustrate the limitations occurring when 29 relying on InSAR and/or PSI measurements for the analysis and interpretation of complex 30 landslide scenarios, especially in cases of relevant spatial and temporal heterogeneities of the 31 deformation field. The joint use of InSAR and DIC can deliver a better picture of the evolution of the deformation field, however, not for all displacement components. Since InSAR, PSI and 32 DIC measurements are nowadays routinely used in the framework of local investigations, as 33 well as in regional, national and/or continental monitoring programs, our results are of major 34 35 importance for users aiming at a comprehensive understanding of these datasets in landslide 36 scenarios.

#### 37 Keywords:

38 Landslide, InSAR, Monitoring, Surface deformation, Swiss Alps

### 39 **1. Introduction**

Remote sensing has proven in the last years to be a valid complement to standard in-situ 40 41 methods for the investigation and analysis of geohazards (Tomás and Li 2017). In particular, optical and radar satellite-based imagery provided great advances in the identification, 42 mapping, and quantification of surface changes caused by earthquakes (Jelének and 43 Kopačková-Strnadová 2021), volcanic deformation (Ebmeier et al. 2018), land subsidence 44 45 (Peng et al. 2022), and slope instabilities (Lissak et al. 2020). A prominent technique to measure surface displacements from space is the Synthetic Aperture Radar differential 46 47 interferometry (InSAR). This approach relies on the identification of phase differences between multi-temporal SAR acquisitions (Bürgmann et al. 2000). The analysis of multi-temporal SAR 48 49 datasets with specific algorithms (e.g., Persistent Scatterer Interferometry, PSI, or Small-Baseline Interferometry, SBAS) allows the generating of ground velocity maps and 50 51 displacement time series, reaching (in ideal cases) sub-centimetric accuracies (Ferretti et al. 52 2001, 2011; Berardino et al. 2002; Werner et al. 2003; Hooper 2008).

53 PSI applications revolutionized the investigation of surface deformation associated with landslide processes (Crosetto et al. 2016). The possibility to obtain data of surface deformation 54 at relatively high spatial and temporal resolutions without installing costly instrumentation is an 55 essential monitoring tool to investigate and interpret landslide processes (Casagli et al. 2016). 56 The current availability of regional, country-scale, and even continental-scale PSI datasets 57 (Crosetto et al. 2020; Lanari et al. 2020) changed the perspective not only in research activities, 58 but also the daily work of practitioners, as well as civil protection strategies (Dehls et al. 2019; 59 Raspini et al. 2019; Bianchini et al. 2021). Intrinsic limitations, however, might hinder the 60 61 nominal performance of InSAR and PSI in mountain areas (Wasowski and Bovenga 2014; 62 Manconi et al. 2018). For example the presence of vegetation and/or snow cover, as well as steep areas affected by geometric distortions in radar geometry cannot be efficiently monitored 63 64 (Cigna et al. 2014). Additionally, InSAR measurements are in one dimension only, i.e., along the satellite's line of sight (LOS), and a combination of a minimum of two or more satellite orbits 65 66 is required to extract displacement in two and three dimensions (Delbridge et al. 2016; Li et al. 67 2019). Due to typically nearly-polar orbits of SAR satellites, however, slope movement along 68 the satellite's track (i.e., almost in North-South directions) remains unmeasured or severely 69 underestimated (Wasowski and Bovenga 2014). Atmospheric phase screen may also seriously affect the accuracy of measurements if not properly considered and corrected (Dini et al. 2019). 70

Another important limitation of InSAR is that accurate measurements are prevented when the surface deformation is relatively large, rapid, and/or spatially and temporally heterogeneous

73 (Manconi 2021). This can be of particular importance when analyzing large landslide

74 complexes, which may have a heterogeneous evolution characterized by non-steady velocities, development of different compartments, and generate potentially local failure events 75 (Stead and Eberhardt 2013; Agliardi et al. 2020). Large and/or rapid displacements can still be 76 77 measured from SAR images with other approaches, as for example the Digital Image 78 Correlation method (DIC, known also as pixel-offset, speckle- or feature-tracking). This is an 79 efficient workaround to retrieve measurements also where large spatial gradients occur; 80 however, the accuracy of the measurements is related to the ground sampling distance (GSD) of the imagery used, which in the case of SAR satellites is typically on the order of meters 81 (Casu et al. 2011; Manconi et al. 2014). Some authors have shown how joint analysis and 82 integration of standard InSAR, DIC, and PSI can be used to investigate large compound 83 84 landslides, although such comprehensive studies are unusual (Singleton et al. 2014; Li et al. 2019). 85



Figure 1: (top) Overview of the Brienz landslide complex (location shown in the inset), background Google Earth imagery. The main morphological distinctions are identified in red (Rutschung Berg, RB) and yellow (Rutschung Dorf, RD), respectively. Two exemplary pictures are shown to highlight the morphological differences between the higher and the lower portions of the slope affected by surface deformation (picture locations and fields of view are shown in the top panel in black).

87 Here we present an extensive analysis of C-Band SAR datasets (frequency 5.4 GHz, 88 wavelength ca. 5.6 cm) acquired from the European Space Agency (ESA) radar missions, i.e., ERS-1/2, Envisat ASAR, and Sentinel-1, covering the period 1992-2020. We reconstruct the 89 90 spatial and temporal evolution of the surface displacements over ca. 30 years at the 91 Brienz/Brinzauls landslide complex (hereafter referred to as Brienz, see Figure 1), located in 92 canton Graubünden (Switzerland). We compare our results against ground-based GNSS 93 measurements and independent information derived from additional radar sensors (Radarsat-94 2 and ALOS-2 PALSAR-2). To complement the InSAR analysis, we also considered DIC on multi-temporal Digital Terrain Models (DTM) generated form airborne LiDAR surveys. The 95 latter were combined with InSAR results to compute the 3-D surface deformation field. 96

97 We completed our analysis up to 2020, i.e. before a further dramatic increase in landslide velocity led to the evacuation of the village of Brienz at the beginning of May 2023, culminating 98 with a failure event on June 15, 2023, with a mass wasting of about 1.2 Mm<sup>3</sup> (Loew et al., 99 100 2023). We deliberately focus on the analysis of the long-term evolution of surface displacements occurring before this major event. The time frame before the event (2020-2023) 101 and additional details on the catastrophic failure are not covered here and will be the subject 102 of future publications, for which specific analyses are presently carried out. The aim of this 103 104 work is manifold. We target at showing and discussing how spatial and temporal 105 heterogeneities might strongly influence the investigation of displacements in complex 106 landslide scenarios when relying on C-Band InSAR satellite measurements only. In most cases 107 only little or no constraints from field data is available; thus, our results in one of the best monitored landslides in the Alps provide important assessments on accuracy and challenges 108 of such analyses in complex scenarios in other regions. In addition, we show how a careful 109 110 use of InSAR based results, not only on PSI but also on wrapped, unwrapped, and stacked interferograms, can be of great value, but often not sufficient to fully judge spatial and temporal 111 heterogeneities. 112

### **113 2. Background on the study area**

The Brienz Mountain slope deformation, located in canton Graubünden, Switzerland, affects a large portion of the southern flank of Piz Linard and the Albula river, and includes the village of Brienz/Brinzauls (765'048 E, 1'170'830 N, CH1903/LV03). The active parts of this mountain slope deformation occur the lower slope areas (below about 1800 m a.s.l and form a very heterogeneous landslide complex. The geomorphological, hydrogeological, and subsurface geological conditions of the active part of this landslide complex have been investigated in detail during the last years (Figi et al., 2022). The currently active upper slope portions (namely

the "Rutschung Berg", i.e., expression to indicate "Landslide affecting the mountain" in German 121 language, hereafter referred to as RB) are located between approximately 1'770 m and 1'150 122 m a.s.l., with an average slope of 36 degrees. The currently active lower slope portions (namely 123 124 "Rutschung Dorf", i.e., expression to indicate the "Landslide affecting the village" in German 125 language, hereafter referred to as RD) extend from about 1'150 m a.s.l. to the Albula riverbed 126 located at approx. 870 m a.s.l, with an average slope of 8 degrees. The landslide involves low-127 grade sedimentary units of the Penninic and Austroalpine nappes with strong differences in mechanical properties. In particular, the RB and RB domains include (from bottom up) located 128 in North Penninic Flysch units, South Penninic Allgäu Schists, Austroalpine Arlberg Dolomite 129 and Raibler Schists units. The Brienz landslide complex is composed of a series of stacked 130 131 old rockslides, rock mass fall deposits and dry/moist granular flows. In the frontal 400-500 m the RD have been thrusted over Late-glacial and Holocene fluvial deposits of the Albula river. 132 Remarkable morphological and geological heterogeneities in RB lead to a distinction into 133 different landslide compartments having different surface velocity amplitudes and directions, 134 as well as different kinematic and dynamic behaviors. In Figure 2, we show the approximate 135 delimitation and naming of the main landslide compartments. A full description of the 136 characteristics of the is beyond the scope of this paper. Details on the geological, structural 137 setting, hydrogeology and dynamic behavior can be found in (Figi et al. 2022). 138



*Figure 2:* Approximate delimitation and naming of the active compartments (cf. Figi et al., 2022). Black dots show the location of GNSS stations (full set with coordinates reported in the Supplementary Information, Table S1). Shaded relief generated from the SwissALTI3D digital model (source, swisstopo, 2013, GSD 2 m).

Large surface displacements and catastrophic failure potential cause major concerns to the 139 140 authorities because approx. 100 people permanently reside in the Brienz village, with up to 200 people during holiday periods. In addition, several important connecting roads and a 141 142 railway line are affected by damages. Slope instability and rock falls have been a recurrent 143 problem for the Brienz population. In the eastern side of the slope, a large earthflow (known 144 as "Igl Rutsch") accelerated in November 1878, and reached velocities of up to 1 m/day, 145 alarming the inhabitants for several months before resting (Ludwig, 2011). Since 1921 the displacements on the village have been periodically measured, and in 2011 a permanent total 146 station was installed to continuously monitor target points in RB and RD. Average surface 147 velocities increased in the entire slope since 2015, and for this reason, the monitoring network 148 149 has been expanded. This includes a combination of periodic surveys and permanent GNSS stations with currently more than 80 points, as well as more than 40 reflectors. Moreover, 150 151 surface displacements are also continuously monitored with a permanent ground-based SAR, and the network is complemented with time lapse cameras acquiring multi-temporal pictures 152 form different locations, as well as with a doppler radar system aimed at detecting rock falls 153 potentially reaching the road connecting Brienz to Vazerol (Schneider et al. 2023). In addition, 154 the dynamic response of the landslide is monitored by analyzing ambient vibration recorded 155 156 through a network of broadband seismometers (Häusler et al. 2022). All these data have been extremely important for the implementation of an early warning system, which allowed to 157 158 recognize timely the critical slope acceleration in spring 2023 and the subsequent evacuation 159 of the population before the slope failure event (Loew et al., 2023).

### **3. Surface deformation from remote sensing**

161 In this work, we considered multiple datasets and methodologies, including:

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(1) Differential interferograms considering all available ESA C-Band satellite imagery (993 163 images in total, see details in Table 2) acquired from ascending and descending orbits of ERS-164 1/2 (1992-2000), Envisat ASAR (2004-2010), and Sentinel-1 (2015-2020). The InSAR analysis 165 166 is performed to obtain initial surface velocity maps by considering different time intervals and 167 served to potentially identify the initiation and the variability of surface displacement in the 168 different domains of the Brienz slope, as well as possibly their main periods of activity. Moreover, we stacked selected Sentinel-1 interferograms from three different orbits, one 169 ascending and two descending. This approach has been used to increase the signal-to-noise 170 ratio, mitigate atmospheric disturbances, and to provide averaged, spatially continuous surface 171 velocity maps over the entire Brienz slope. 172

Satellite	Orbits (# images)	Period available	Revisit time (days)
ERS-1/2	T215A (#38) T480D (#65)	1992-2000	35
Envisat ASAR	T215A (#37) T480D (#42)	2004-2010	35
Sentinel-1	T015A (#258) T066D (#275) T168D (#278)	2015-2020	6*

*Table 2.* Summary table of the ESA C-band SAR dataset available. The A in the orbit name stands for ascending tracks, while D for descending. \*The Sentinel-1 revisit time of 6 days is available after 2017. Before this date the revisit time is 12 days.

174 (2) Persistent Scatterer Interferometry (PSI) on data acquired from ERS-1/2, Envisat
175 ASAR, and Sentinel-1, each on ascending and descending orbit. We considered the
176 Interferometric Point Target Analysis (IPTA, Werner et al., 2003), aiming at computing surface
177 velocity maps and displacement time series at point targets maintaining a good coherence
178 (i.e., SAR signal quality).

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180 (3) Digital Image Correlation (DIC) applied to four Digital Terrain Models (DTMs) generated
181 from airborne LiDAR. The data were acquired by helicopter on November 12, 2015, June 15,
182 2018, December 6, 2019, and September 4, 2020, respectively. The DIC analysis provided
183 measurements over areas affected by large displacements, i.e., where DInSAR usually fails to
184 provide useful results.

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(4) 3-D surface displacements for two selected periods, i.e., 2015-2018 and 2018-2020.
Sentinel-1 stacking computed in (1) and the DIC results obtained in (3) have been jointly used
to obtain a full description of the displacement field, as well as hints on the temporal evolution
of the landslide complex.

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To ease following our workflow and the reading, in the sections below we sequentially describe
the methods and directly present the most relevant results. Additional outcomes (in particular,
selected wrapped interferograms) are reported in the Supplementary Information (see Figures
S1-S8) and discussed in section 5.

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# **3.1. Differential interferometry and multitemporal stacking.**

199 We computed surface displacement maps with differential interferometry by considering pairs of SAR images after alignment (also known as co-registration) and removal of the phase 200 contribution due to topography (Bürgmann et al. 2000). The products obtained (i.e., differential 201 interferograms) were visually inspected in their wrapped phase form, and then unwrapped by 202 considering the MCF algorithm (Costantini 1998). For each satellite and orbit, we stacked 203 selected interferograms with good quality (i.e., spatial coherence). Stacking methodologies are 204 frequently used in geophysical data processing and remote sensing analyses in order to 205 improve the signal-to-noise ratio (Stumpf et al. 2017; Gorelick et al. 2017). Regarding InSAR, 206 207 stacking is generally performed by combining multiple unwrapped differential interferograms covering a pre-defined time period (e.g., one or multiple years), in order to highlight spatial 208 209 domains retaining surface velocities within this time frame (Lundgren et al. 2001; Ciuffi et al. 210 2021). Any artefacts affecting single acquisitions are expected to have random variability over 211 time and thus be mitigated with integration and/or averaging of the signal, provided that enough 212 input images are available.

#### 213 3.1.1 Results

No clear signs of surface displacements are evident in the period 1992-2010 with standard 214 InSAR analyses on ERS-1/2 and Envisat ASAR (see also Supplementary Information S1-S2), 215 although during this period minor surface activity was already known and measurable in-situ. 216 Single interferograms considering temporal baselines ranging from 35 days to one year were 217 not successful in retrieving suitable measurements due to the excessive noise levels. Failure 218 to retrieve measurements can be explained mainly due to relatively poor quality (in terms of 219 spatial and temporal baselines) of the data available in this time range in comparison to the 220 rates of motion and landcover. Figure 3 shows the results of annual stacking obtained for the 221 222 Sentinel-1 orbits, which provide the most valuable results when selecting interferograms with 223 perpendicular baselines (orbit separations) below 150 m and temporal baselines as short as 224 12 days. Interferograms with longer temporal baselines suffer from a substantial drop in 225 coherence and were not beneficial in our investigation.



Figure 3: Results of the annual Sentinel-1 stacking analysis. Velocities are in satellite LOS. Negative values (blue) mean that the ground moved away from the satellite, while positive values (red) indicate that the ground moved towards the satellite. Black solid lines represent the main connection roads and are shown in all tiles to provide a spatial reference. Missing results are due to layover/shadow masking.

Sentinel stacks were generated considering the same observation periods for each orbit, consistently starting in the beginning of September of each year, highlighting inter-annual variability. We set coherence thresholds aiming at increasing the quality of the final stacking products, however, no significant improvement was observed and thus we decided to present the results here without thresholding. We attempted shorter stacking, covering about three months to potentially identify intra-annual and/or seasonal variabilities, but results were inconclusive. We also tried stacking solely Sentinel-1 interferograms with temporal baselines
of 6 days, but this approach drastically reduced the number of input images, compromising the
effectiveness of the procedure.

A striking observation are the remarkable differences in the spatial distribution of the 235 236 displacement signal when comparing ascending and the two descending results. Results of 237 T015A highlight mainly LOS movement away from the satellite in RB and the outer extents of RD in a "horseshoe-like" pattern. Distinct signals showing a different LOS sign for RB and RD 238 239 are instead detected in descending orbits, where the surface displacements change sign even 240 within the RD domain at the landslide toe, in the area close to the Albula river. This spatial 241 variability suggests a strong heterogeneity of the surface displacements directions, combined with the remarkable topographic variations of the area under investigation. Focusing on the 242 temporal behavior, a progressive increase of the surface displacements can be observed at 243 selected locations, although this is not very clear in all considered orbits (Figure 4). For 244 245 example, surface velocities at the Brienz village appear to be very small in the period 2016-2017 compared with the portions of RB located at higher elevations. After 2017, the RD 246 displacement pattern starts to be more evident and locally of the same order of magnitude as 247 RB. The upper extent of RB also shows a remarkable increase in surface velocities over the 248 249 analyzed time. Near the GNSS stations 6006 and 6001 (see locations in Figure 2), LOS surface 250 velocities were of about 0.2-0.4 m/a in the period 2016-2018 but reached values up to 1 m/a 251 in 2019-2020. Moreover, eastern portion of the instability shows a remarkable acceleration 252 from 0.2 m/a in 2016-2017 to about 0.3-0.4 m/a in 2018-2020. Some discrepancies between the results obtained by different orbits can be related to the changes in viewing angles and/or 253 to random noise on the stacking products. Considering the variability of the results in the 254 vicinity of areas assumed stable over the analyzed period, for example near GNSS point 42 255 located in the village of Vazerol, the expected level of accuracy of the stacking results is on 256 the order of 0.1 m/a. This value is higher in areas where the displacements are of several m/a. 257

## **3.2 Interferometric Point Target Analysis (IPTA)**

PSI is considered an advanced remote sensing method; however, in the last few years this approach has gained more and more popularity and can be considered as a standard approach for the investigation of surface displacements on unstable slopes (Wasowski and Bovenga 2014). This is mainly due to the availability of new generation satellites such as the ESA Sentinel-1 constellation, which provides reliable acquisitions at global scales and with unprecedented spatial and temporal resolutions (Torres et al. 2012).



*Figure 4:* Results of the annual stacking at selected locations (see Figure 2 for the position of points). Velocities are in satellite LOS.

Several projects have been developed to provide wide-area coverage of PSI results (Zinno et 265 al. 2018; Dehls et al. 2019; Crosetto et al. 2020), which directly provide to the end-users PSI 266 results in the form of surface velocity maps and displacement time series. Here we performed 267 268 PSI processing on two relevant Sentinel-1 tracks, by applying the IPTA method on "single-269 reference" stacks (GAMMA Software, Wegmüller et al., 2016). This is one of the most diffused 270 approaches for multi-temporal DInSAR analyses over unstable slope regions. The reference images for the IPTA processing were selected in the middle of the entire time of analysis (about 271 five years), i.e., July 23, 2018, and July 27, 2018, for T015A and T066D, respectively. This 272 choice helped to balance the temporal baselines of the processed pairs and expected to 273 274 improve interferometric coherence. Due to the accurate control of the Sentinel-1 orbital parameters, the spatial baseline between all acquisitions and the reference image is generally 275 below 150 m, ensuring a good quality of the SAR phase and thus more reliable displacement 276 measurements. 277

#### 278 3.2.1 Results

Figure 5 shows the Sentinel-1 surface velocities over the Brienz slope complex achieved with IPTA processing, compared with previous results obtained by considering ERS and Envisat ASAR sensors in the ESA GMES TERRAFIRMA project (Raetzo et al. 2007), for ascending and descending orbits. The relatively poor spatial coverage of point measurements is as expected. Some signs of surface displacements on the order of 0.01 cm/a can be seen in the

village of Brienz starting from the period 2002-2010 (Figure 5c), however, as already 284 285 evidenced in the standard InSAR and stacking analyses, the surface deformation starts to be relevant only when considering the Sentinel-1 results. The measurement points are located 286 287 mainly at buildings and/or other anthropic infrastructures, as well as some rock faces 288 maintaining a relatively good temporal coherence over the period of analysis. While no 289 information is generally available from the IPTA on the RB portions, measurement points 290 located at the Brienz village appear to move towards the satellite LOS in descending orbit and 291 away from the satellite LOS in ascending orbit. This suggests that a relevant component of the 292 surface displacement affecting the Brienz village is slope parallel and/or sub-horizontal. In the 293 Sentionel-1 results (Figure 5e-f), some isolated points can be seen southwest of Brienz village 294 that show an opposite sign of surface velocity. This suggests the presence of local areas with different displacement directions; however, the overall number of IPTA points is too small to 295 296 perform a comprehensive interpretation. Snow cover in winter periods and rapid seasonal changes deeply affect the phase correlation in alpine regions (Wasowski and Bovenga, 2014). 297 This cannot be avoided despite performing local investigations on the slope of interest, and 298 299 thus, with all processing parameters calibrated to obtain the best solution over the study area. 300 The time series mostly show a linear displacement trend, with some minor oscillations that might be related to seasonal variations and/or uncompensated atmospheric artefacts (see 301 Supplementary Information, Figure S9). The ground-based measurements from 2015-2020 a 302 303 substantial acceleration in the same time period (Figi et al. 2022). Discrepancies between IPTA 304 results and in-situ observations are likely caused by the large and rapid displacements 305 occurring over the Brienz slope complex (Manconi 2021).

# **306 3.3. Digital Image Correlation analysis**

Digital Image Correlation (DIC) is a method allowing to track displacements by comparing pixel 307 groups in multi-temporal digital imagery acquired from different sensors, mainly optical and 308 309 radar, and from different platforms (ground-based, airborne, and/or spaceborne). The DIC 310 strategy has been initially adopted on remote sensing datasets to study rapid flow of glaciers (Strozzi et al. 2002; Kääb et al. 2009). More and more often, however, this technique is used 311 312 for the investigation of slope instabilities and landslide events, despite a reduced accuracy compared to standard InSAR approaches (Manconi et al. 2014). The theoretical accuracy of 313 slope displacements measured by DIC strongly depends on the signal-to-noise ratio of the 314 imagery, as well as on the GSD of the input data (Bickel et al., 2018; Bontemps et al., 2018; 315 Stumpf et al., 2017). For this reason, Sentinel-1 imagery (with GSD of about 3 m in range and 316 15 m in azimuth direction) is often not suitable to obtain the desired accuracy levels at the 317 318 slope scale.

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Figure 5: Surface velocity map over the Brienz slope obtained by applying the IPTA approach on different C-Band satellite sensors. Positive values (red) mean that the ground moved towards the satellite LOS, while negative values (blue) indicate that the ground moved away from the satellite LOS. (a) ERS ascending, 1992-2000, note the poor spatial coverage due to reduced amount of imagery in comparison with the descending orbit; (b) ERS descending, 1992-2000; (c) Envisat ASAR ascending, 2002-2010; (d) Envisat ASAR descending, 2002-2010; (e) Sentinel-1 ascending, 2015-2020; (f) Sentinel-1 descending, 2015-2020. Selected Sentinel-1 PS time series are presented in the Supplementary Information, Figure S9.

Here we considered multi-temporal DTMs generated from airborne LiDAR surveys over the 320 321 Brienz slope in the period between 2015 and 2020. First, we resampled the input data to a common grid of 1 m, as they initially entailed different GSDs. Then, we generated shaded relief 322 323 images from the DTMs by considering multi-directional sun azimuth and elevation (Fey et al. 324 2015). The shaded relief data is projected coordinates in map coordinates (CH1903, EPSG 21781), thus, DIC results provide measurements of the displacement components occurring 325 in the North-South and East-West directions. The DIC processing was performed using the 326 software presented in (Bickel et al. 2018), which has already demonstrated good performances 327 328 for large slope deformation related to other alpine mass movements (Manconi et al. 2018; Glueer et al. 2019; Storni et al. 2020; Aaron et al. 2021). We have done several tests to identify 329 the parameters providing reliable results. The final DIC parameters used in this analysis are 330 reported in the Supplementary Information, Table S2. 331

#### 332 3.3.1 Results

333 DIC results are presented for two selected time windows considered important for the interpretation of the spatial and temporal evolution of the Brienz slope complex. The periods 334 335 range between November 12, 2015, and June 15, 2018 (period 1), and between June 15, 2018, and September 4, 2020 (period 2). These time windows have similar durations (2.6 vs. 336 2.2 years) and are characterized by the transition from slow to moderate surface velocities 337 (i.e., period 1) and then a period of sustained high velocities (i.e., period 2). The results (Figure 338 S10) show the dominating displacement direction towards South, with velocity values 339 exceeding 2 m/a especially in the upper RB portions as well as along the western extents of 340 RB. The East-West displacement component is lower in magnitude, and for this reason the 341 DIC results include more noise. Another important observation from the DIC results is that the 342 spatial distribution of the surface displacements is very similar in the two considered periods, 343 however, an overall increase in surface velocities is noted in the 2018-2020 period, indicating 344 an acceleration compared to the 2015-2018 period. This agrees with the ground-based 345 observations. One of the main issues in the DIC processing was related to boundary effects 346 associated with poor coverage of the DTMs around the Brienz slope, i.e., in areas expected to 347 348 have minor or no displacements and that can be used for the evaluation of the DIC accuracy.

## 349 3.4. Reconstruction of 3-D displacements field

350 Due to the increased data availability, the integration of InSAR results from multiple satellite 351 orbits, as well as with DIC products, is increasingly popular in the analysis of slope deformation.



*Figure 6. 3-D surface velocities in m/a over the Brienz landslide complex, obtained by combining the Sentinel-1 stacking and the DIC results on the LiDAR DTMs. The black dashed polygon shows the area where minor uplift has been identified (see text for more details).* 

Using this approach it is possible to derive a 2-D and/or 3-D representations of the 352 displacement field and better study the kinematic behavior of landslides (Elefante et al. 2014; 353 Delbridge et al. 2016; Frattini et al. 2018; Crippa et al. 2020). In the specific case of Brienz, a 354 straightforward application of the classical approach combining the Sentinel-1 ascending and 355 descending orbits does not provide substantial benefits. The main reason is that the SAR 356 357 satellites fly on a near-polar orbit (~11 degrees from North), and thus all surface displacements 358 oriented along the satellite's track are not measurable because they produce only very small 359 (if any) changes in the LOS. Indeed, the displacement components that can be retrieved from 360 two or more orbits are 2-D, i.e. the vertical and East-West directions, assuming that the North-361 South component is negligible (Manzo et al. 2006). From the DIC results, as well as from the ground-based monitoring, we know that in Brienz the displacement amplitudes show relevant 362 spatial variations, and that the main component of slope movement is oriented towards South. 363 To reconstruct the 3-D displacement components, an integration of all the available results is 364 necessary. We computed the 3-D deformation considering a least-squares estimation of the 365 displacement components starting from multiple observations. The least-squares solution is 366 constructed as an over-determined set of linear equations, with the design matrix solved into 367

the Singular Value Decomposition (see for example (Casu and Manconi 2016). We exploited in total five observations (three LOS measurements from the Sentinel-1 orbits and two DIC results, in North-South and East-West directions, respectively). To this end, we have resampled the input datasets to a common grid (5x5 meters) and computed stacking products to align the temporal observation of the Sentinel-1 data with the LiDAR surveys. The calculation was thus done on the two reference periods 2015-2018 and 2018-2020.

#### 374 3.4.1 Results

Figure 6 shows the 3-D surface velocities for the time periods 2015-2018 and 2018-2020. The 375 results confirm the dominance of the South-oriented motion component. The increase of 376 surface velocity in the period 2015-2018 can be recognized in all components, however, the 377 378 changes are more remarkable in the North-South and Up-Down directions, where the velocity locally exceeds 2 m/a in RB. Uplift of few cm/a was identified West of the Brienz village, with 379 380 an increasing trend in the 2018-2020 period (~6 cm/a) compared to 2015-2018 (~2 cm/a). Unfortunately, no monitoring points were installed in this area during the period of observation, 381 382 thus we could not compare these results with external data. Additional GNSS survey and 383 levelling measurements were started in summer 2021 to better understand the behavior of this zone. The source of this uplift is not clear with the information currently available. 384

#### **4. Comparison with GNSS measurements**

The reliability and accuracy of the surface deformation obtained with the combination of InSAR 386 387 and DIC analyses has been validated with the available GNSS in-situ measurements. The 388 extensive network of stations across the Brienz landslide complex allows for validation of the 389 deformation results obtained in remote sensing analyses. We extracted the surface velocities 390 for each direction at the location of GNSS stations and compared them with the velocities 391 recorded by GNSS over the same time periods. We considered that GNSS station 42 is our stable reference (see location in Figure 2). Figure 7 shows the scatterplots comparing the 392 stacking results obtained for the three different Sentinel-1 orbits with the GNSS measurements 393 projected along the satellite LOS. The agreement is better in the 2015-2018 period, when 394 surface velocities were relatively lower than during 2018-2020. However, the inaccuracy of the 395 InSAR measurements increases as the surface velocity overcomes the temporal phase 396 aliasing thresholds at 0.85 m/a, considering 6-days revisit time (0.425 m/a with 12-days revisit 397 time). This is more pronounced for the points located in the RB and leads to an underestimation 398 of surface velocities up to several m/a in comparison with GNSS measurements. 399 400



*Figure 7.* Scatterplots comparing surface velocities measured with Sentinel-1 InSAR stacking and in-situ GNSS (projected along satellite LOS). Perfect agreement is indicated when points lie on the plots' diagonal. Red dashed lines show the theoretical phase aliasing limits for 6 and 12 days (see Manconi, 2021).



*Figure 8.* Scatterplots comparing 3-D surface velocities obtained with the combination of Sentinel-1 InSAR stacking and DIC versus in-situ GNSS. Perfect agreement is indicated when points lie on the plots' diagonal. Red dashed lines show the theoretical phase aliasing limits (see Manconi, 2021).

401 Figure 8 shows the comparison of GNSS with the 3-D surface velocities obtained by combining the Sentinel-1 stacking and the DIC results. The agreement improves compared to the results 402 403 of the InSAR stacking only, however, large discrepancies can still be observed. This is more 404 evident in the vertical components, as the DIC results do not contribute to the least squares 405 solution, and in the East-West components, as motion in these directions is relatively small 406 compared to the North-South component and mostly below the DIC detection thresholds (see also Supplementary Information, Figure S10). In general, the main discrepancies with the 407 GNSS measurements are related to the underestimation of the vertical component in the upper 408 portions of the RB. The North-South component gains better agreement to the GNSS 409 410 measurements; however, in the 2018-2020 period there is still a clear underestimation of the surface velocities. 411

412

# 413 **6. Discussion and conclusions**

The results attained with extensive investigations of remote sensing datasets can be of great value for the evaluation of the state of activity of landslide complexes. However, we have shown that several problems can be encountered. The main difficulties arise during processing of the critical stage from slow to fast surface displacement observed at large deep-seated slope instabilities (Agliardi et al. 2020). This period can last for several years or rapidly evolve on the order of few weeks and potentially increase the probability of catastrophic slope failures.

420

421 We provide numerous indications of the spatial and temporal evolution of surface displacement 422 occurring at the Brienz slope complex in the period 1992-2020, ca. 30 years. First, we revealed 423 that PSI analyses over heterogeneous, complex landslide scenarios such as Brienz, affected 424 by large and/or rapid surface velocities, are not suitable for a comprehensive investigation of the spatial and temporal kinematic evolution. The main problems are related to the reduced 425 coherence of the SAR phase, which leads to inaccurate displacement measurements and poor 426 PS spatial coverage. The analysis of selected Sentinel-1 interferograms indicates that the RB 427 domain in the period 2015-2018 has constant high levels of surface displacements, which can 428 hardly be accurately detected with InSAR of image pairs acquired more than 18 days apart 429 due to phase decorrelation. By stacking Sentinel-1 interferograms, it has been possible to 430 reconstruct at least the yearly trends of surface displacements on one ascending and two 431 descending orbits. These results provide a fair overview of the spatial distribution of surface 432 deformation over the last five years, as well as information on their temporal evolution. In the 433 434 RB domain the displacement amplitudes have increased between 2015 and 2020, while the

RD domain shows an overall similar displacement behavior throughout the analyzed period.
The only area with a relevant increase in surface displacement within this time frame is located
to the West of the Brienz village.

438

439 We have also used additional satellite sensors to cross-validate the InSAR results. The 440 analysis of the Radarsat-2 dataset has provided suitable results for a limited number of 441 interferograms covering the period 2014-2019 despite the very-high spatial resolution of 5 m (see also Supplementary Information S3-S4). Moreover, the information is available only for 442 443 summer periods. This period is already covered by the Sentinel-1 imagery, which has a more frequent revisit time compared to Radarsat-2 (6 to 12 days vs. 24 days) and thus provided 444 445 better results (see also Supplementary Information S5-S7). ALOS-2 PALSAR-2 satellite data 446 (L-band SAR imagery) for selected dates can retrieve more deformation than C-band in the 447 RB area (see also Supplementary Information S8). Due to its longer wavelength (ca. 23 cm), L-band is less impacted by phase decorrelation caused by large movement and potentially 448 local failure, and/or can follow displacements occurring in areas with vegetation (Aoki et al. 449 450 2021). Additional indications are provided on the surface deformation affecting the areas are 451 not covered by GNSS or total station monitoring (eastern slope portions). Unfortunately, the L-452 band data were very limited over our area of interest. Additional datasets acquired from SAOCOM, as well as future satellite missions planned with L-band SAR such as the NISAR, 453 454 ALOS-3 PALSAR-4 and ROSE-L, will be an important source of information for landslide 455 analyses in complex scenarios such as the Brienz case (Rosen et al. 2017).

456

457 Due to the intrinsic limitations in spatial resolution and in retrieving North-South displacement components with InSAR, the results obtained from the Sentinel-1 stacking analysis alone 458 cannot be conclusive. We have thus performed a DIC analysis by considering other remote 459 460 sensing data, i.e., the DTMs obtained from airborne LiDAR surveys. This investigation is complementary to InSAR, as it provides information on displacements along the East-West 461 and especially North-South directions. Considering the ground resolution of the available 462 DTMs, the accuracy of the DIC analysis is expected to be on the order of  $\pm 0.1$  m/a (Bickel et 463 al., 2018), which is suitable to provide information in the specific case of Brienz, where the 464 465 surface velocities are of m/a. The comparison between two selected sub-periods, i.e., 2015-466 2018 and 2018-2020, shows a clear increase in surface velocities mainly in the RB domain 467 and towards the South. However, slight accelerations are also observed in the RD domain, 468 especially in the westernmost portions at the elevations of the Brienz village. This is one of the areas where major damage is witnessed on the roads and infrastructures. Another area where 469 470 we observe a slight increase in surface velocities is on the Southeast sectors towards the

Albula river. DIC approaches are a valid complementary method for the analysis of surface displacement components; however, datasets with suitable GSD are necessary to retrieve accurate measurements. The case of Brienz is rather unique, as it is unusual to have multitemporal LiDAR surveys within 5 years. In the future, very high-resolution imagery available from new generation satellites (optical and SAR) with a spatial resolution on the order of 1 m will be an important additional source of information to be considered in such investigations.

We have combined the results of DInSAR stacking and the DIC analysis to reconstruct a full, 478 479 3-D picture of the surface deformation affecting the Brienz landslide complex. The combination of multiple Sentinel-1 orbits and DIC results has provided important information on the spatial 480 distribution of the surface deformation. This approach allowed us to retrieve a more accurate 481 representation of the displacement field and improved the agreement with the GNSS 482 measurements. The surface deformation is characterized by a "patchy" pattern, which is 483 representative of internal landslide compartments. This is in agreement with detailed 484 evaluations of the 3-D displacements performed for the RB area and based on high resolution 485 486 terrestrial LiDAR datasets (Kenner et al. 2022). The most relevant observations on the 3-D 487 surface velocities are related to the upper reaches of the RB domain (i.e., North of GNSS station 6001), where the surface velocities have substantially increased during the period 488 2018-2020 compared with 2015-2018. The obtained 3-D results show that intrinsic limitations 489 490 associated with phase aliasing can be only partially overcome. In the North-South direction, 491 where the displacements are of several m/a, we could retrieve a better agreement between 492 with the GNSS measurements. This is because the DIC analysis provides a good complement 493 to the InSAR measurements. East-West components, however, are smaller in amplitude and thus suffer of accuracy problems also with DIC. On the other end, up-down components are 494 exclusively measured through InSAR and cannot be mitigated with this analysis. Higher 495 resolution datasets or constellation of satellite SAR sensors with more frequent acquisitions 496 can help in retrieving a better representation of the 3-D behavior and allow for a more detailed 497 kinematic interpretation of landslides based on the analysis of the displacement vector 498 orientations (Kenner et al. 2022). 499

500

Among the discussed limitations, we note that a substantial underestimation of the landslide acceleration can occur if only spaceborne C-Band (i.e., currently Sentinel-1) interferometry is applied. Indeed, phase decorrelation might affect the spatial sampling of measurement points, while phase aliasing can cause large deviations from the correct velocity values. This has to be sensibly considered in systematic monitoring programs aimed at identifying surface displacement anomalies (Dehls et al. 2019; Crosetto et al. 2020; Bianchini et al. 2021). The

507 parameters determining the accuracy of the results when using spaceborne and aerial imagery 508 are the spatial and temporal resolution, as well as their intrinsic detection thresholds. This seems to be a trivial concept; however, it is not always carefully considered, and interpretations 509 510 can be misleading. The amount of information available at the Brienz site is exceptional, and only in few alpine sites remote sensing investigations can be complemented with detailed 511 surface geology and structural data, and/or validated with in-situ measurements. Especially in 512 513 inaccessible regions, where remote sensing is often the only available data source to define 514 active landslide domains and possibly build evolutionary scenarios and/or hazard assessments, we suggest a cautious evaluation of the InSAR results obtained with C-Band 515 516 sensors only.

517

# 518 7. Authors' Responsibilities

AM and SL conceived the study. NJ performed the Sentinel-1 InSAR and stacking processing. AM performed the IPTA analysis, DIC, 3-D combination, and the validation with GNSS. TS performed the processing of Radarsat-2 and ALOS-2 PALSAR-2 interferometry. RC and UW performed the pre-processing of the Sentinel-1 datasets. AM, NJ and SL interpreted the results. AM and NJ wrote the paper. All co-authors revised the manuscript.

# 524 8. Declaration of Interest

525 The authors declare no conflict of interest.

## 526 9. Acknowledgements

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# 706 Supplementary Information

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Table S1. Coordinates (in Coordinate System CH1903/LV03, EPSG 21781) of the GNSS
 points reported in Figure 2. The reported date is referred to the last measurement considered
 in this work.

	Х	Y	Z	Lasi
GNSS Point	Coordinate	Coordinate	Coordinate	
11	764635.52	171058.8	1165.95	18.08.20
12	764630.68	170634.28	1073.81	18.08.20
13	764680.33	170430.26	1021.68	18.08.20
14	764693.69	169952.06	898.11	18.08.20
15	764527.54	169863.47	872.9	18.08.20
22	765034.81	170380.71	1004.44	18.08.20
23	765120.08	169995.9	904.92	18.08.20
31	765230.36	171434.87	1331.13	28.05.18
32	765459.98	170910.46	1112.86	18.08.20
33	765553.79	170571.24	1008.5	18.08.20
34	765672.72	170285.44	923.04	23.05.13
35	765794.15	170077.26	885.04	18.08.20
41	763874.68	170972.07	1125.15	14.06.19
42	764131.97	170843.6	1134.77	14.06.19
114	764704.98	169987.35	907.16	18.08.20
115	764523.38	169862.71	872.64	18.08.20
122	765057.1	170386.28	1003.55	18.08.20
123	765122.15	169997.69	905.09	14.06.19
131	765316.51	171464.36	1346.68	28.05.18
134	765674.22	170284.59	922.8	18.08.20
1001	764654.03	171653.21	1573.18	11.05.20
1002	764657.92	171741.34	1602.53	11.05.20
1003	764650.44	171819.48	1644.93	11.05.20
1004	764690.84	171848.46	1669.81	29.03.19
1005	764774.18	171868.76	1697.39	11.05.20
1006	764808.18	171882.92	1713.96	24.05.17
1007	764806.59	171975.58	1752.97	11.05.20
1008	764798.25	171978.35	1755.04	11.05.20
1009	764788.01	171967.76	1753.23	11.05.20
1010	764831.6	172072.77	1790.05	11.05.20
1011	764845.87	172164.24	1796.07	14.06.19
1012	764742.94	172199.98	1813.08	14.06.19
1013	764817.08	172329.05	1791.34	14.06.19
1014	765145.54	172294.12	1760.12	14.06.19
1015	765280.08	172039.36	1694.16	14.06.19
1016	764598.13	171537.24	1487.05	11.05.20
1017	764595.99	171540.78	1487.56	03.11.17
2001	764695.8	171899.54	1701.93	11.05.20
2002	764712.44	171972.57	1734.2	11.05.20
2003	764723.83	172073.76	1776.56	11.05.20
2004	764706.81	172123.57	1803.5	11.05.20

2005	764849 87	172025 3	1765.08	11 05 20
2006	764947 99	171954 64	1712 27	11.00.20
2000	764960 2	171994.39	1741 12	11.00.20
2008	764939 27	172050 1	1763.04	11.05.20
3001	764122 73	171726 65	1438.6	14 06 19
3002	764078.37	171292.87	1285.96	14.06.19
3003	763830.86	171329.31	1222.76	14.06.19
3004	764358.14	170368.38	1004.29	14.06.19
3005	765407.69	171035.03	1162.84	11.05.20
3006	765597.03	171193.68	1214.38	14.06.19
3007	765636.04	171401.77	1369.73	14.06.19
3008	765879.13	171825.06	1484.39	14.06.19
3009	765802.9	171946.3	1537.51	14.06.19
3010	765991.29	172043.78	1539.14	14.06.19
3011	765479.64	172117.39	1646.6	14.06.19
3012	765687.8	172298.1	1717.9	14.06.19
3013	765100.9	172559.88	1861.83	14.06.19
3014	764612.36	172563.9	1881.27	14.06.19
3015	764559.12	172269.29	1763.02	14.06.19
3020	764556.48	170753.96	1114.53	18.08.20
3021	764357.72	171051.89	1200.88	18.08.20
3022	764276.44	170954.53	1187.21	18.08.20
3023	764295.98	171308.81	1304.29	18.08.20
3024	764329.59	171318.34	1309.42	18.08.20
5001	764993.2	170891.21	1157.19	04.11.20
5002	764793.98	170585.19	1064.86	04.11.20
5003	765476.83	170674.25	1054.6	04.11.20
6001	764518.48	171419.19	1378.04	04.11.20
6002	764649.29	171819.72	1644.96	04.11.20
6003	764797.02	171968.55	1753.15	04.11.20
6004	764866.93	171970.95	1738.31	04.11.20
6005	764702.34	171888.14	1698.95	04.11.20
6006	764703.79	171977.2	1736.68	04.11.20
7001	764422.86	171221.12	1286.39	04.11.20
7002	764447.31	170892.64	1156.6	04.11.20
12160151	763927.61	170397.7	1012.92	14.06.19
12160218	765443.67	172557.38	1851.03	14.06.19
12160224	764120.6	171722.74	1438.87	14.06.19
12160229	765889.63	170914.1	1093.29	14.06.19
12160230	765035.5	170709.18	1127.26	18.08.20
12160235	764594.81	170121.85	940.21	18.08.20

- Table S2.Summary table of the parameters used for the Digital Image Correlation analysis.Additional information on the parameters and their effect on the results can be found in
- Bickel et al., 2018.

Parameter	Value
Template window (pixels)	128x128
skip (pixels)	16
Oversampling factor	2

Figure S1. Selected interferograms for the ERS1/2 and Envisat ASAR satellite imagery, ascending orbit. The results are strongly affected by phase decorrelation and do not allow us to identify whether surface deformation was already present over the Brienz landslide complex during the period 1992-2010. 



Figure S2. Selected interferograms for the ERS1/2 and Envisat ASAR satellite imagery, descending orbit. The results are strongly affected by phase decorrelation and do not allow us to identify whether surface deformation was already present over the Brienz landslide complex during the period 1992-2010.





Figure S3. Selected interferograms for the Radarsat-2 satellite imagery, ascending orbit. The
 results are strongly affected by phase decorrelation.





- Figure S4. Selected interferograms for the Radarsat-2 satellite imagery, descending orbit.
- 745 The results are strongly affected by phase decorrelation, however some interferograms show
- 746 deformation over RB and RD.



- 749
- 750

Figure S5. Selected interferograms for the Sentinel-1 satellite imagery, ascending orbit T015.



Figure S6. Selected interferograms for the Sentinel-1 satellite imagery, descending orbitT066.



Figure S7. Selected interferograms for the Sentinel-1 satellite imagery, descending orbitT168.



Figure S8. Selected ALOS PALSAR-2 L-band interferograms, descending orbit. (top)
Comparison between 14-days interferogram in L-band and 6-days interferogram in C-band.
(bottom) Comparison between 14-days and 3-months interferograms with L-band, detail over
RB and IL.



20200910 – 20200924 ALOS-2 Descending orbit

20200914 – 20200920 Sentinel-1 Descending orbit



20200910 – 20200924 ALOS-2 Descending orbit

 $-\pi$ 



π



**Differential Phase** 

Sentinel-1, rad @5.4GHz (28 mm) ALOS-2, rad @1.3GHz (115 mm)

Figure S9. Selected Sentinel-1 ascending (top) and descending (bottom) displacement time series results for a point located in the Brienz village. Displacements are in the satellite LOS. The opposite displacement directions, i.e., away from satellite in ascending orbit and towards the satellite in descending orbit, denotes a strong horizontal component of the deformation







Figure S10. (top) DIC results for the East-West direction (unfiltered). The largest surface velocities are in the RD domain. In some areas, a slight increase is noted in the 2018-2020 period compared to 2015-2018. Some local artefacts are visible in the results due to rock fall activity (black circles). (bottom) DIC results for the North-South direction (unfiltered). The largest surface velocities are in the RB domain. There, a remarkable increase is noted in the 2018-2020 period compared to 2015-2018. Some local artefacts are visible in the results due to rock fall activity (black circles) and road maintenance (black rectangle).



(West) -0.5 < \_\_\_\_ > 0.5 (East)





