

1 Dear Editor and Referees,

2 We wish to thank you for your effort in improving our paper. We agree that several parts of the
3 manuscript needed to be reviewed. In particular we feel that now this work is more incisive and its
4 purpose clearer. This was thanks to your suggestions, especially those relative to cutting unnecessary
5 parts or to better explaining some others.

6 Following, we resume all the answers to your requests together in this single file because they are tightly
7 linked to each other and we think that in this way they are more understandable.

8 In order to give informed answer, we preferred to modify the manuscript according to your revisions,
9 although this was not necessary at this stage (this is why we could not upload it yet). In this way, we
10 were able to answer your suggestions not just theoretically but according to what we actually changed in
11 the manuscript.

12 Best regards on behalf of all co-authors,

13 Emanuele Intrieri

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Answers to S. L. Gariano (Editor)

Dear Authors, both reviewers considered your work good and publishable, after revisions. In particular, they believe that the “technical” parts of the work should be reduced and/or summarized. I agree with them. Thus, I have also some suggestions, listed below:

- Text in rows 61-83 could be largely reduced.

It has been almost entirely deleted (see also answers to Referee #1 and #2).

- Sections 3.1, 3.2, and 3.3 could be reduced.

Section 3.1: the part relative to how interferometric data are made and how they are elaborated has been deleted (see also answers to Referee #1 and #2).

Section 3.2: this has been reduced and a figure has been deleted (see also answers to Referee #1).

Section 3.3: this has been reduced as suggested.

- Figures 4 and 5 could be deleted.

They have been deleted from section 4. For reference, in the new version their numbering was no more 4 and 5 but 3 and 4.

- Finally, I suggest reviewing the abstract by adding more precise information about the method described in the paper and the obtained results.

Sincerely, Stefano Luigi Gariano

Thank you for your observation. In fact we probably missed to properly convey the message of our paper. This might have created confusion in some of the comments of the two referees concerning the fact that our aim and obtained results are not the monitoring data themselves, rather the procedures employed to obtain them. Therefore we changed the final part of the abstract and replaced it with the following sentence:

“The aim of this paper is to show how logistic issues linked to advanced monitoring techniques such as big data transfer and storing, can be dealt with, compatibly with an early warning system. Therefore, we focus on the interaction between an areal monitoring tool (a ground-based interferometric radar) and the DCPC. By converting complex data into ASCII strings and through appropriate data cropping and average, and by implementing an algorithm for line of sight correction, we managed to reduce the data daily output without compromising the capability of performing”.

44 **Answers to Anonymous Referee #1**

45 General comments

46 The paper presents a procedure for the integration of GB-InSAR data within an early warning integrated
47 system for risk prevention for a critical infrastructure (A16 highways connecting Naples to Bari in southern
48 Italy). The use of GBInSAR for landslide monitoring is not new in the scientific literature, although not yet
49 standardized, so its use in EWS is certainly of interest to the community of landslide researchers. The used
50 language is correct and readable. However, some changes are suggested before publication on NHES
51 journal. The weak point is that the current version of the paper appears as a “technical note” rather than an
52 original research article. Indeed, the Authors provide plenty of details concerning the LEWIS system (that is
53 not central in the work) and both the installation and set-up of GB-InSAR tool but, on the other hand, the
54 interpretation of the results and the adoption of thresholds for early warning purposes based on GB-InSAR
55 data is not given the same room and relevance. So, given the timely topic addressed and the potential of
56 these kind of applications, Authors are invited to better balance different parts of the paper to improve the
57 overall quality and readability of the work for the typical audience of NHES. Some suggestion are provided
58 hereafter.

59 As the Editor highlighted, technical note are not available for this Journal. The paper has been balanced
60 following the suggestions furnished in the following comments of the reviewer. In particular, we have
61 better explained the method used here to set thresholds but, since the setting of thresholds is not the
62 objective of the paper, we have also explained that the system is open and different methods can be
63 implemented as well. Furthermore, the part concerning LEWIS has been reduced and a figure removed, in
64 order to better balance the topic of the paper.

65

66 Specific comments

67 Lines 55 to 83 provide too many details anticipating the technical descriptions that are expected instead in
68 section 3 or 5. Please remove from here.

69 This paragraph has been in part moved in paragraph 5 and in large part deleted since it mainly anticipated
70 concepts more deeply described in paragraph 5.

71 In section 3.1 the description of LEWIS should be reduced since the Authors already refer to the published
72 work of Costanzo et al. (2016).

73 The section containing information about LEWIS (3.2) has been reduced as suggested and in particular a
74 figure has been removed (Fig. 2). Only the parts that are important to allow the reader to easily understand
75 the following sections have been kept.

76 In section 6, please better clarify how GB-InSAR data interpretation and analysis contribute to fix thresholds
77 for early warning.

78

79 Technical corrections: - In the abstract do not use future tense (line 29) –

80 Done

81 Lines 76, 78: change “where” in “were”.

82 This part has been removed.

83 - Figure 1: change the shaded fonts because they are not readable

84 The font has been changed and the shaded box now have a solid colour.

85 - Figure 2: increase the font size.

86 This figure has been removed.

87 - Lines 251, 252: use the past tense.

88 Done.

89 - Line 327: add references to:

90 - Cascini et al., 2010 (for first maps of DInSAR data projected along the steepest slope direction). Cascini L.,
91 Fornaro G., Peduto D. (2010). Advanced low- and full-resolution DInSAR map generation for slowmoving
92 landslide analysis at different scales. Engineering Geology, 112 (1-4), 29-42,
93 doi:10.1016/j.enggeo.2010.01.003.;

94 and to Cascini, L., Peduto, D., Pisciotta, G., Arena, L., Ferlisi, S., and Fornaro, G. (2013): The combination of
95 DInSAR and facility damage data for the updating of slow-moving landslide inventory maps at medium
96 scale, Nat. Hazards Earth Syst. Sci., 13, 1527-1549, doi:10.5194/nhess-13-1527-2013, for the map of
97 projectable DInSAR data.

98 The references to Cascini et al. have been added.

99 - Line 412: please clarify better to which "friction" you are referring.

100 We were referring to the friction between vehicles and the tar. Now it is specified in the text.

101

102

Answers to Anonymous Referee #2

103 The authors submitted a work to present the role of GB-InSAR on an integrated system for landslide
104 monitoring. In particular, the early warning system architecture and the data management are treated. The
105 work is referred to the LEWIS project and it is focused on a critical infrastructure in southern Italy (A16
106 highways). Although the technology here presented is now well-known, the integration with other
107 monitoring technique and the development of EWS are interesting topics. The objectives of the manuscript
108 are clear and the paper covers an area of interest to the journal's readership.

109 In order to improve the manuscript, I recommend authors to summarize the section with irrelevant details
110 for readers (eg. GB-InSAR features..) focusing on data analysis and interpretation, also by adding some
111 displacement time series.

112 The sections highlighted by the referee have been heavily reduced and 3 figures have been removed in
113 total. Concerning the time series, we have not included them since these are not really meaningful. In fact,
114 the slope did not experience significant movements and the growth of the vegetation produces noise that
115 concealed the slightest deformations, which did not exceed the instrumental resolution (less than one
116 millimeter per day). On the other hand, displacement maps (now figure 7) have been included, in order to
117 show some displacement data nonetheless and to show that displacements were negligible. Indeed, as you
118 rightly pointed out, the use of GB-InSAR for landslide monitoring is not new and it was not the aim of our
119 paper. Our scope was to explain a procedure to overcome some logistic issues encountered in an early
120 warning system, such as big data. We have added to better explain our scope in the abstract and in the
121 introduction. They are reported below.

122 In the abstract: "The aim of this paper is to show how logistic issues linked to advanced monitoring
123 techniques such as big data transfer and storing, can be dealt with, compatibly with an early warning
124 system. Therefore, we focus on the interaction between an areal monitoring tool (a ground-based
125 interferometric radar) and the DCPC. By converting complex data into ASCII strings and through
126 appropriate data cropping and average, and by implementing an algorithm for line of sight correction, we
127 managed to reduce the data daily output without compromising the capability of performing."

128 In the introduction: "One of the main drawbacks of advanced instruments such as GB-InSAR is how to
129 handle the large data flow deriving from continuous real-time monitoring. The issue is to reduce the
130 capacity needed for analyzing, transmitting and storing big data without losing important information. The
131 main feature of this paper is indeed the management of monitoring data in order to filter, correct, transfer
132 and access them compatibly with the needs of an early warning system."

133 In addition, it is really important to improve the conclusions, also by focusing on data integration for EWS.

134 Thank you for pointing out this issue. We improved the conclusions by better explaining the possible
135 usefulness of our paper with reference to similar situations. Unfortunately, we do not have data from other
136 instruments; in fact, all the monitoring devices were independent and the integration was only needed at a
137 higher level, when monitoring data and results from modeling were finally integrated and a risk assessment
138 was possible. These aspects are already treated in other paper cited in the manuscript (Versace et al., 2012;
139 Costanzo et al., 2015; 2016). In this paper, we only deal with the interaction between GB-InSAR and the
140 DCPC. In fact, what happen next (e.g. data integration between GB-InSAR and other instruments) falls out
141 of our interest and detailed knowledge.

142 Following, some specific comments and minor points to improve the text:

143 - Page 2, row 56: please replace "aerial" with "spatial"

144 Done.

145 - Page 2, rows 56-57: please replace "The installation was in an area where the only internet connection
146 available 57 was 3G" with "the monitoring area was covered by a 3G mobile telecommunication networks"

147 This sentence has been changed as suggested and it is now at the beginning of section 5, following the
148 suggestion of referee #1.

149 - Page 2, rows 61-83: these lines are very specific and of little interest to readers. Please consider deleting
150 these lines or inserting them in Section 3.3.

151 They have been deleted and only in small part moved.

152 - Page 2, row 76: please change “where” with “were”

153 This part has been removed from the paper.

154 - Page 3, row 86: please change "ground-based interferometric synthetic aperture radar" with "Ground-
155 Based Interferometric Synthetic Aperture Radar (GB-InSAR)"

156 We have now changed this.

157 - Section 3.1: this section appears too long. Please consider reducing sentences and adding a table with the
158 technical specifications of equipment used (eg. Frequency, Bandwidth, Range and Cross-range resolution,
159 etc.).

160 **Thank you for this comment. In fact this part was mostly a repetition and has been largely reduced. We**
161 **also remove the part explaining the technical specifications of the equipment, since this information can**
162 **be found in the literature cited in the paper and is not fundamental for our purpose.**

163 - Section 4: please add more geological information (eg. materials involved) to better frame the area under
164 study.

165 **Now we have explained that “The lithologies outcropping in this area are Pliocene-Quaternary clay,**
166 **clayey marlstones, and more recent (Holocene) terraced alluvial sediments (from clay to gravel). The**
167 **landslides shown in Figure 2 are all located in clay or clayey marlstones”.**

168 - Page 8, row 250: please add space before "These"

169 **Done.**

170 - Fig. 1: please improve the quality of figure

171 **Now the text is bolder and the boxes are no longer filled with a gradient but with a solid color.**

172 - Fig. 3: please increase the font size

173 **The font size has been increased.**

174 - Fig. 8: please insert the location of GB-InSAR instrument installed.

175 **You are right. The yellow asterisk in the left of the images represents the location of the GB-InSAR. This is**
176 **now specified in the caption of the image.**

177

178 **Big data managing in a landslide Early Warning System: experience from a ground-based**
179 **interferometric radar application**

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189

190

191 **Keywords:** early warning system; slope instability; big data; monitoring; landslide; risk
192 management; ground-based interferometric radar

193

194 **1 Abstract**

195 A big challenge in terms of landslide risk mitigation is represented by the increasing of the
196 resiliency of society exposed to the risk. Among the possible strategies to reach this goal, there is
197 the implementation of early warning systems. This paper describes a procedure to improve early
198 warning activities in areas affected by high landslide risk, such as those classified as Critical
199 Infrastructures for their central role in society.

200 This research is part of the project “LEWIS (Landslides Early Warning Integrated System): An
201 Integrated System for Landslide Monitoring, Early Warning and Risk Mitigation along Lifelines”.

202 LEWIS is composed of a susceptibility assessment methodology providing information for single
203 points and areal monitoring systems, a data transmission network and a Data Collecting Aand
204 Processing Center (DCPC), where readings from all monitoring systems and mathematical models
205 converge and which sets the basis for warning and intervention activities.

206 The aim of this paper is to show how logistic issues linked to advanced monitoring techniques such
207 as big data transfer and storing, can be dealt with, compatibly with an early warning system. In this
208 paperTherefore, we ~~will~~ focus on the interaction between an areal monitoring tool (a ground-based
209 interferometric radar) and the DCPC. By converting complex data into ASCII strings and through
210 appropriate data cropping and average, and by implementing an algorithm for line of sight
211 correction, we managed to reduce the data daily output without compromising , and how issues
212 such as big data transfer, the capability of performing real-time warning, line of sight correction and
213 data validation in emergency conditions have been dealt with.

214

215 2 Introduction

216 Urbanization, especially in mountain areas, can be considered a major cause for high landslide risk
217 because of the increased exposure of elements at risk. Among the elements at risk, important
218 communication routes, such as highways, can be classified as Critical Infrastructures (CIs), since
219 their rupture can cause chain effects with catastrophic damages on society (Geertsema et al 2009;
220 Kadri et al. 2014). On the other hand, modern society is more and more dependent from CIs and
221 their continuous efficiency (Lebaka et al., 2016), and this has risen their value over the years. The
222 result is a higher social vulnerability in the face of loss of continuous operation (Kröger, 2008). The
223 main objective was to improve the social preparedness to the growing landslide risk, according with
224 the suggestions of several authors (Gene Corley et al., 1998; Baldrige et al., 2011; Urlainis et al.
225 2014; 2015). This led to the development of several approaches and frameworks for increasing the
226 resiliency of society exposed to the risk (Kröger, 2008; Cagno et al., 2011 and references therein).
227 The resiliency policy of course involves prevention activities but also, and more importantly, those
228 activities needed to maintain functionality after disruption (Snyder and Burns, 2009) and to
229 promptly alert incoming catastrophes in order to protect people and prepare for a possible damaging
230 of the endangered CI. Among these activities, the implementation of integrated landslides early
231 warning systems (*i.e.* LEWIS, Versace et al., 2012; Costanzo et al., 2016) reveals its increasing
232 importance.

233 In this context, the methodology described in this paper has been conceived; it has been tested and
234 validated on a portion of an Italian highway, affected by landslides and selected as case study: it is
235 located in Southern Italy, along a section of the A16 highway, an important communication route
236 that connects Naples to Bari where a ground based interferometer (GB-InSAR) has been installed
237 on the test site, in order to obtain aerialspatial monitoring data.

238 One of the main drawbacks of advanced instruments such as GB-InSAR is how to handle the large
239 data flow deriving from continuous real-time monitoring. The issue is to reduce the capacity needed
240 for analyzing, transmitting and storing big data without losing important information. The main
241 feature of this paper is indeed the management of monitoring data in order to filter, correct, transfer
242 and access them compatibly with the needs of an early warning system.-

243 ~~A ground based interferometer (GB-InSAR) has been installed on the test site, in order to obtain~~
244 ~~aerial monitoring data. The installation was in an area where the only internet connection available~~
245 ~~was 3G, with a limit of 2 gigabyte data transfer per month. Nevertheless, these data could be~~
246 ~~managed thanks to the implemented data transmission network and Data Collecting and Processing~~
247 ~~Center (DCPC), organized taking into account both the internet network problems and the big~~
248 ~~amount of data produced by the interferometer.~~

249 ~~Interferometric data are indeed complex numbers, organized in a matrix where each pixel contains~~
250 ~~both phase and amplitude information of the backscattered signal (Bamler and Hartl, 1998;~~
251 ~~Antonello et al., 2004); the radar employed produced a 1001x1001 complex matrix (corresponding~~
252 ~~to ~7 megabytes) every 5 minutes. Therefore, there was the need to reduce the massive data flow~~
253 ~~produced by the radar. For this reason data were locally and automatically elaborated in order to~~
254 ~~produce, from a complex matrix, a simple ASCII grid containing only the pixel by pixel~~
255 ~~displacement value, which is derived from the phase information. Then, since interferometry only~~
256 ~~measures the displacement component projected along the radar line of sight, data needed to be re-~~
257 ~~projected. This was performed by dividing the ASCII grid by a correction matrix, where every~~

258 element of the matrix was the percentage of the actual displacement that was measurable by the
259 radar; such percentage can be obtained with trigonometrical arguments knowing the position of the
260 radar and the direction of movement of the landslides (which, in our case, corresponded with the
261 slope direction) thus enabling the calculation of the radar line of sight.

262 To further reduce the size of the grids, matrixes were cropped in order to contain only those pixels
263 where relevant information could be extracted.

264 The ASCII grids were also averaged to reduce noise, so 8 hours and 24 hours averaged grids were
265 obtained. According to the early warning procedures that were defined, during periods characterized
266 by low or null slope movement, only 8 hours and 24 hours data were transferred, together with the
267 last displacement measurement of a reduced number of control points.

268 The transfer was performed after transforming the grids into strings and by sending them through a
269 middleware to the Data Acquisition and Elaboration Centre, where control points displacement
270 values were compared with warning thresholds and the grids were projected on a GIS
271 environment as 2D displacement maps.

272

273 3 Materials and methods

274 3.1 GB-InSAR

275 The **g**round-**b**ased **I**nterferometric **S**ynthetic **a**perture **R**adar (GB-InSAR) is composed of a
276 microwave transceiver mounted on a linear rail (Tarchi et al., 1997; Rudolf et al., 1999; Tarchi et
277 al., 1999). The system used is based on a Continuous Wave – Stepped Frequency radar, which
278 moves along the rail at millimeter steps, in order to perform the synthetic aperture; the longer the
279 rail the higher the cross-range resolution. The microwave transmitter produces, step-by-step,
280 continuous waves around a central frequency, which influences the cross-range resolution and
281 determines the interferometric sensitivity i.e. the minimum measurable displacement, usually
282 largely smaller than the corresponding wavelength.

283 The installation was in an area where the only internet connection available was 3G, with a limit of
284 2 gigabyte data transfer per month. Nevertheless, these data could be managed thanks to the
285 implemented data transmission network and Data Collecting and Processing Center (DCPC),
286 organized taking into account both the internet network problems and the big amount of data
287 produced by the interferometer.

288 Interferometric data are indeed complex numbers, organized in a matrix where each pixel contains
289 both phase and amplitude information of the backscattered signal (Bamler and Hartl, 1998;
290 Antonello et al., 2004); the radar employed produced a 1001x1001 complex matrix (corresponding
291 to ~7 megabytes) every 5 minutes. Therefore, there was the need to reduce the massive data flow
292 produced by the radar. For this reason data were locally and automatically elaborated in order to
293 produce, from a complex matrix, a simple ASCII grid containing only the pixel by pixel
294 displacement value, which is derived from the phase information. Then, since interferometry only
295 measures the displacement component projected along the radar line of sight, data needed to be re-
296 projected. This was performed by dividing the ASCII grid by a correction matrix, where every
297 element of the matrix was the percentage of the actual displacement that was measurable by the
298 radar; such percentage can be obtained with trigonometrical arguments knowing the position of the

299 ~~radar and the direction of movement of the landslides (which, in our case, corresponded with the~~
300 ~~slope direction) thus enabling the calculation of the radar line of sight.~~
301 ~~To further reduce the size of the grids, matrixes where cropped in order to contain only those pixels~~
302 ~~where relevant information could be extracted.~~
303 ~~The ASCII grids where also averaged to reduce noise, so 8 hours and 24 hours averaged grids were~~
304 ~~obtained. According to the early warning procedures that were defined, during periods characterized~~
305 ~~by low or null slope movement, only 8 hours and 24 hours data where transferred, together with the~~
306 ~~last displacement measurement of a reduced number of control points.~~
307 ~~The transfer was performed after transforming the grids into strings and by sending them through a~~
308 ~~middleware to the Data Acquisition and Elaboration Centre, where control points displacement~~
309 ~~values where compared with warning thresholds and the grids where projected on a GIS~~
310 ~~environment as 2D displacement maps.~~

311 The radar produces complex radar images containing the information relative to both phase and
312 amplitude of the microwave signal backscattered by the target ([Bamler and Hartl, 1998](#); [Antonello](#)
313 [et al., 2004](#)). The amplitude of a single image provides the radar reflectivity of the scenario at a
314 given time, while the phase of a single image is not usable. The technique that enables to retrieve
315 displacement information is called interferometry and requires the phase from two images. In this
316 way, it is possible to elaborate a displacement map relative to the elapsed time between the two
317 acquisitions.

318 The main added value of GB-InSAR is its capability of blending the boundary between mapping
319 and monitoring, by computing 2D displacement maps in near real-time. The use of this tool to
320 monitor structures, landslides, volcanoes, sinkholes is largely documented ([Calvari et al., 2016](#); [Di](#)
321 [Traglia 2014](#); [Intrieri et al., 2015](#); [Bardi et al., 2016, 2017](#); [Martino and Mazzanti, 2014](#); [Severin,](#)
322 [2014](#); [Tapete et al., 2013](#)), as well as for early warning and forecasting ([Intrieri et al., 2012](#); [Carlà et](#)
323 [al., 2016a; 2016b](#); [Lombardi et al., 2016](#)).

324 GB-InSAR systems probably reveal their full potential in emergency conditions. They are
325 transportable and only require from few tens of minutes to few hours to be installed (depending on
326 the logistics of the site). Moreover, they ~~are able to can~~ detect "near-real time" area displacements,
327 without accessing the unstable area, 24h and in all weather conditions ([Del Ventisette et al., 2011](#);
328 [Luzi, 2010](#); [Monserrat et al., 2014](#)). On the other hand, some limitations reduce the GB-InSAR
329 technique applicability: first of all the scenario must present specific characteristics in order to
330 reflect microwave radiations, maintaining high coherence values ([Luzi, 2010](#); [Monserrat et al.,](#)
331 [2014](#)); only a component of the real displacement vector can be identified (i.e. the component
332 parallel to the sensor's line of sight); maximum detectable velocities are connected to the time that
333 the system needs to obtain two subsequent acquisitions. Sensors need power supply that, for long
334 term monitoring, cannot be replaced by batteries, generators or solar panels.

335 With the specific aim of performing an early warning system, data acquired *in situ* must be sent
336 automatically to a "control center" where they are integrated in a complete early warning system
337 procedure ([Intrieri et al., 2013](#)). In this sense, another main limitation is represented by the necessity
338 to transfer a high quantity of data, whose weight has to be reduced to the minimum, in order to
339 reduce the load on transmission network.

340 The employed system is a portable device designed and implemented by the Joint Research Center
341 (JRC) of the European Commission and its spin-off company Ellegi-LiSALab (Tarchi et al., 2003;
342 Antonello et al., 2004). ~~The linear rail is 210 cm long, allowing a synthetic aperture of maximum
343 180 cm. It is easily transported by a normal motorized vehicle but its length is enough to obtain
344 high cross resolution images, also at a distance from the scenario the can reach more than 1 km.
345 The transmitter and receiver move on the rail on a specific support, which can be tilted in order to
346 direct the microwave radiation as much as possible parallel to the displacement direction. The
347 power base represents the control center of the system; it contains a personal computer that
348 manages the definition of input parameters used in the acquisition phase and a UPS (Uninterruptible
349 Power Supply) to guarantee constant electric supply; it also enables data elaboration and storage,
350 being equipped with two boards increasing the memory to 1.8 TiB. The employed system needs 850
351 W, 230 VAC and 50 Hz as power supply. The integrated UPS guarantees the continuity of the
352 electric supply in case of necessity, for a period of maximum 12 hours after the power cut. The total
353 weight of the instrument is 95 kg, equally distributed over the different components (power base
354 with UPS and boards, transeeiver, linear rail).~~

355 *3.2 Early warning system architecture*

356 Morphological features, hydrogeological factors and sudden rainfall can cause ~~different~~ diverse
357 types of movements or fall of earthy and rock materials. The unpredictability and diversity of these
358 events make structural interventions often inappropriate to reduce the related risk, and real-time
359 monitoring network difficult to implement.

360 In the last decade, Wireless Sensor Networks (WSNs) have been largely used in various fields. A
361 significant increase in the use of WSN, due to their simplicity, low cost of installation,
362 manufacturing and maintenance, has been recorded in the framework of environmental monitoring
363 applications (Intrieri et al., 2012; Liu et al., 2007; Yoo et al., 2007). Different-Distinct types of
364 sensor nodes of these networks, distributed with high density in the monitored areas, send
365 environmental information to the concentrators nodes, generating a considerable amount and a wide
366 variety of collected data. Due to the significant growth of data volumes to be transferred, the WSN
367 require flexible ad-hoc protocols, able to respect constraints related to energy consumption
368 management (Hadadian and Kavian, 2016; Khaday et al., 2015; Parthasarathy et al., 2015). In
369 particular, many protocols have been developed that offer data aggregation patterns to optimize the
370 sensor nodes battery life (Kim et al., 2015) or sleep/measurement/data transfer cycles to minimize
371 the energy consumption (Fei et al., 2013; Venkateswaran and Kennedy, 2013).

372 LEWIS (Costanzo et al., 2016) uses heterogeneous sensors, distributed in the risk areas, to monitor
373 the several physical quantities related to landslides. The measured data, through a
374 telecommunications network, flow into the Data Collecting Aand Processing Center (DCPC),
375 where, using suitable mathematical models for the monitored site, the risk is evaluated and
376 eventually the state of alert for mitigation action is released (Figure 1).

377 The system, through a modular architecture exploiting a telecommunication network (called
378 LEWARnet) based on an ad-hoc communication protocol and an adaptive middleware, has a high
379 flexibility, which allows for the use of different interchangeable technological solutions to monitor
380 the parameters of interest.

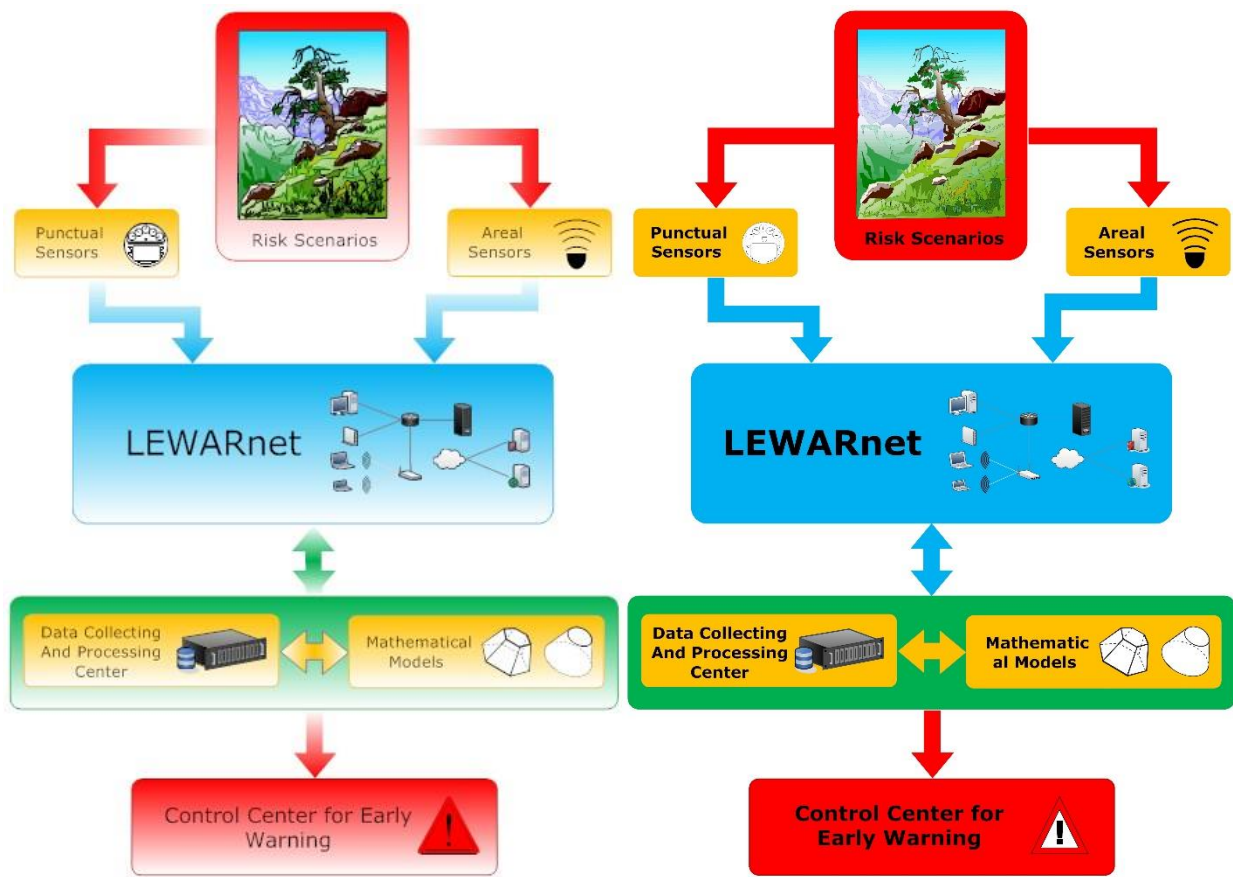


Figure 1. LEWIS architecture.

381

382

383 The telecommunication network has been designed and implemented within the Microwave
 384 Laboratory of the University of Calabria.

385 The areas to be monitored have been divided into geomorphological units, indicated as GU. Each
 386 GU is then subdivided into more heterogeneous sensors subnets.

387 For this purpose, each GU contains a concentrator node, called first level Sink, which has the
 388 purpose of coordinating the sensors or any sub-network of sensors and collect the measurements
 389 from them to transmit the data to the DCPC. Each sub-network is coordinated by a second level
 390 Sink node, connected to the first level Sink through a short-range wireless connection (eg. ZigBee,
 391 Bluetooth).

392 The network architecture can be schematized as shown in Figure 2.

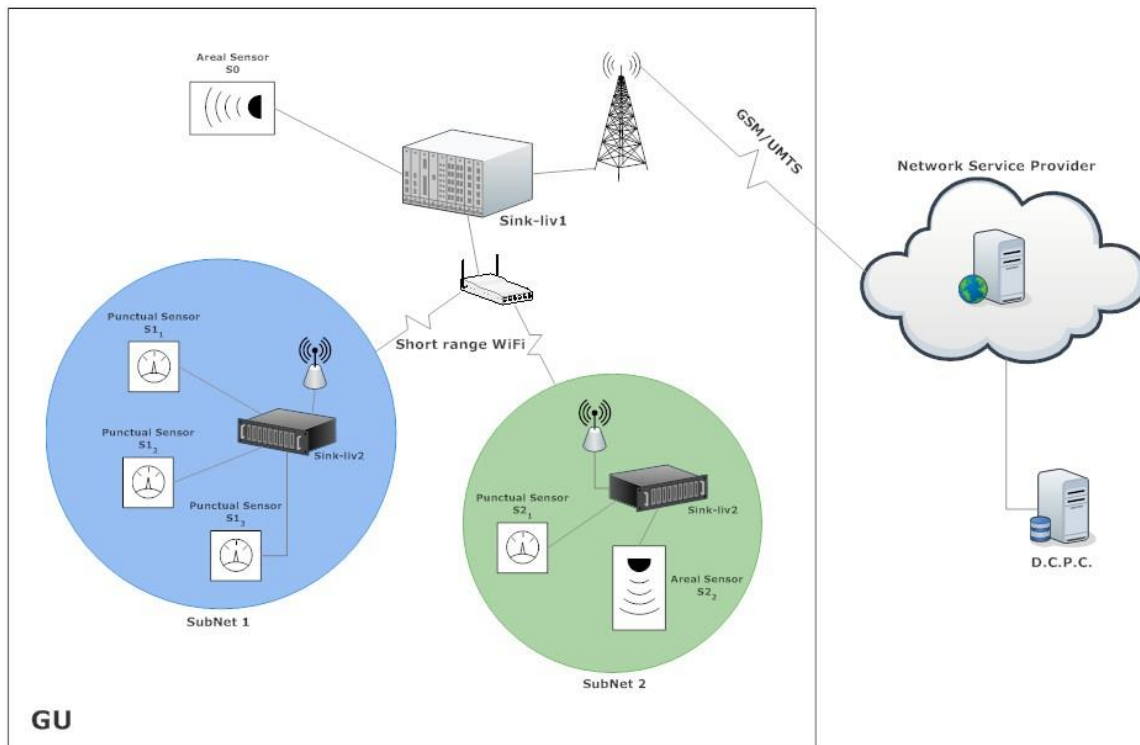


Figure 2. LEWARnet architecture.

393
394

395 The network has been equipped with both single point sensors as well as area sensors, ~~and the sub-~~
396 ~~networks may include the use of a single type of these or both.~~ The present paper addresses a sub-
397 network comprising an area sensor, the GB-InSAR.

398 The different sensors types generate asynchronous traffic, thus imposing the adoption of an ad-hoc
399 transmission protocol. This can support an asynchronous transmission mode to the DCPC, and it is
400 equipped with message queues management capacity to reconstruct historical data series, between
401 two connection sessions, in case of null or partial transmission. This operation mode requires the
402 presence of a software architecture that operates as a buffer, acting as an intermediary or as
403 middleware (LEWARnet), between the data consumer (DCPC) and the data producers (sensors and
404 sub-networks of sensors).

405 The developed middleware also monitors the processes of transmission and data acquisition,
406 recognizing the activity status of the sensors and that of the DCPC, and integrating encryption and
407 data compression functions.

408 [A detail description of LEWIS can be found in Costanzo et al. \(2015; 2016\).](#)

409

410 3.3 Data Collecting and Processing Center (DCPC)

411 The management of information flows, the telematic architecture and the services for data
412 management are entrusted to the DCPC.

413 The DCPC has been designed and performed according to a complex hardware and software
414 system, able to ensure the reliability and continuity of the service, providing advance information of
415 possible dangerous situations that may occur.

416 In the research project, the DCPC has to ensure the continuous exchange of information among
417 monitoring networks, mathematical models and the Command and Control Center (CCC), that is
418 responsible for emergency management and decision making.

~~419 The design and implementation of procedures for the exchange of information from the monitored
420 sites to the CCC was built according to persistent and stable communication protocols, that are
421 suitable for hardware/software architecture of monitoring devices, for models and for CCC.~~

422 Data flow from the monitoring network was managed according to a communication protocol,
423 implemented by the DCPC, and named AqSERV. AqSERV was designed considering the
424 heterogeneity of devices of monitoring and transmission networks (single point and area sensors)
425 and the available hardware resources (microcontrollers and/or industrial computers). AqSERV was
426 devised to link DCPC database (named LEWISDB) to the monitoring networks, after validation for
427 the authenticity of the node that connects to the center. Data acquisition, before the storage in the
428 database, is validated both syntactically and according to the information content. The procedures
429 for extraction of the information content and validation have been realized differently for single
430 point and area sensors: the latter require a more complex validation, as they work in a 2D domain.

431 The complete management of the monitoring networks by DCPC has been realized through specific
432 remote commands, sent to individual devices via AqSERV, to reconfigure the acquisition intervals
433 or to activate any sensor, depending on the natural phenomena occurring in real time.

~~434 The acquired and validated data are then accessible for the mathematical models through a further
435 service, created ad hoc, which publishes all the acquisitions by sensors on a remote server for
436 sharing.~~

437 The configuration of monitoring networks, composed by devices and sensors, of communication
438 protocol used by each network, and of rules for extraction and validation of information content is
439 carried out through a web application that allows for the management of the whole-entire system by
440 the users.

~~441 Besides the configuration, the application has been configured to automatically create the tables of
442 interest; automation of the process permits to reduce the acquisition time and possible human errors.~~

443 The real-time search for acquisitions is carried out through a WebGIS, specifically designed for
444 WSNs, but that can be easily extended to classic monitoring networks.

445 The WebGIS was designed according to the traditional web architecture, client-server, by using
446 network services which are web mapping oriented:

- 447 - web server for static data;
- 448 - web server for dynamic data;
- 449 - server for maps;
- 450 - database for the management of map data.

~~451 The static layers provided by the WebGIS are the results produced by geological studies for the
452 identification of event scenarios: geological map, geomorphological map, map of event scenarios.~~

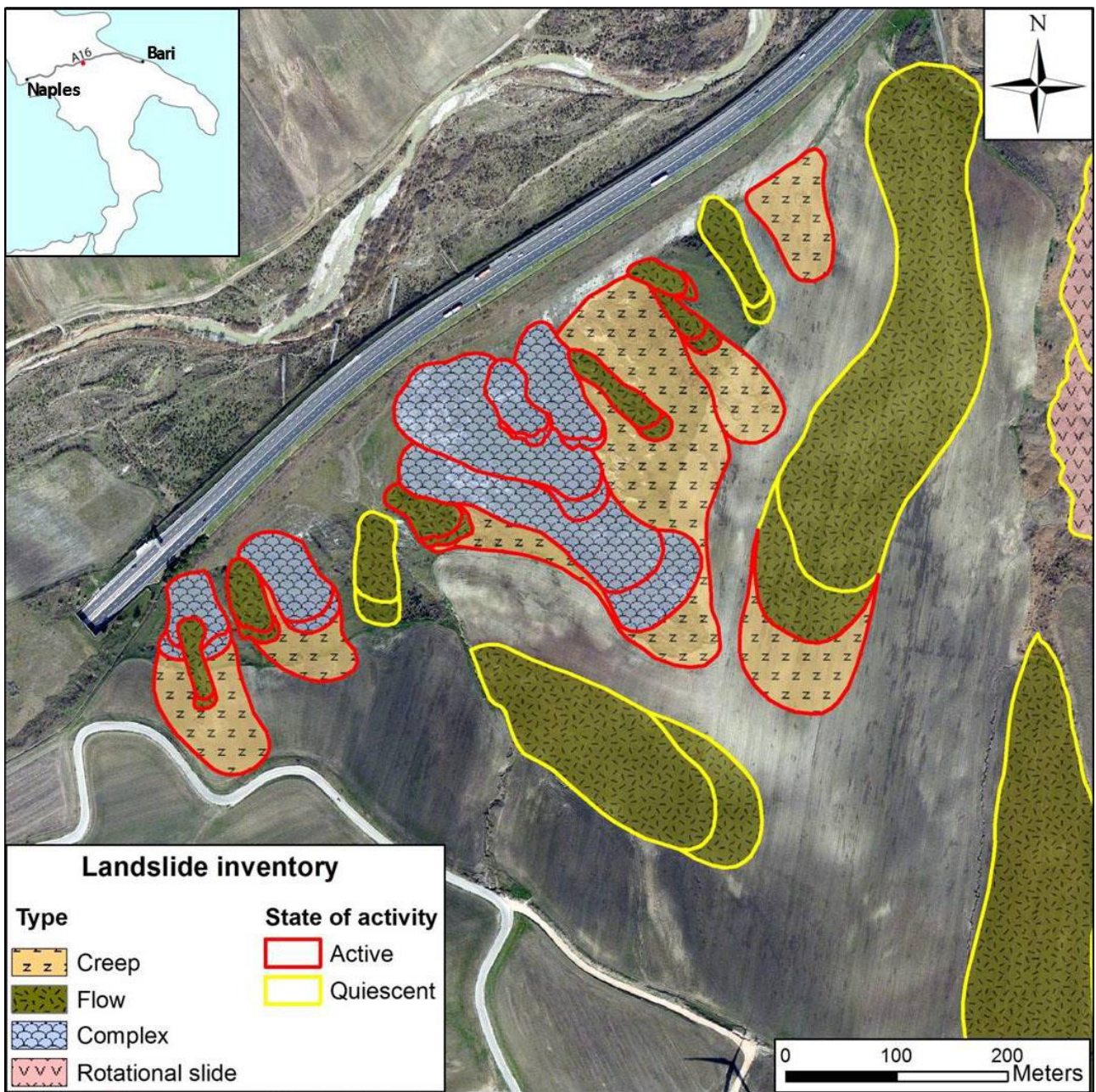
~~453 The dynamic layers are the acquisitions in real time by the sensors.~~

~~454 A DCPC operator can consult the information provided by each layer via a standard web browser
455 verifying the performance of the event precursors and any anomalies in acquisitions.~~

456 **4 Test site**

457 The test site chosen to experiment the integrated system is located in Southern Italy, along a section
458 of the A16 highway, an important communication route that connects Naples to Bari (Figure 2).

459 The A16 selected section develops in SW-NE direction, along the Southern Italian Apennine, in
 460 correspondence with the valley of the Calaggio Creek, between the towns of Lacedonia (Campania
 461 Region) and Candela (Puglia Region).



462
 463 **Figure 2. Landslides detected through field survey along the monitored section of A16 highway.**

464 The area is tectonically active, but the landscape, characterized by gentle slopes, is mostly
 465 influenced by lithologic factors ~~(the strong presence of clayey sediments)~~ rather than by tectonics.
 466 The lithologies outcropping in this area are Pliocene-Quaternary clay, clayey marlstones, and more
 467 recent (Holocene) terraced alluvial sediments (from clay to gravel). The landslides shown in Figure
 468 2 are all located in clay or clayey marlstones.

469 The highway runs on the right flank of the Calaggio Creek at an altitude between 300 and 400 m
 470 a.s.l.; the section of interest represents an element at risk in the computation of landslide risk

471 assessment, due to the presence of unstable areas which can potentially affect the communication
472 route (Figure 2). These unstable areas mainly involve clayey superficial layers.

473 On 1st July 2014, the GB-InSAR system ~~has-was-been~~ installed on the test site. The location of the
474 installation point ~~has-been-was~~ selected taking into account the view of the unstable area and the
475 distance from the power supply network. A covered structure was built ~~in-order~~ to protect the
476 system from atmospheric agents and possible acts of vandalism, in the perspective of a long-term
477 monitoring (~~Figure 3~~).



478

479 **Figure 3. Pictures of the construction of the covered structure and installation of the GB-InSAR.**

480 The transmission network was provided by a GSM modem, exploiting the 3G network. In addition
481 to the PC integrated in the GB-InSAR power base, a further external PC was exclusively employed
482 for data post elaboration and transmission.

483 The system acquired from the beginning of July 2014 until the end of July 2015.

484 The installation location allowed the system to detect an area between 40 and 400 meters far from
485 the its position in range direction, and about 360 m wide in the azimuth direction. These values,
486 coupled with a 40° vertical aperture of the antennas, allowed operators to detect an area of about
487 360 m x 360 m (~~Figure 4~~).



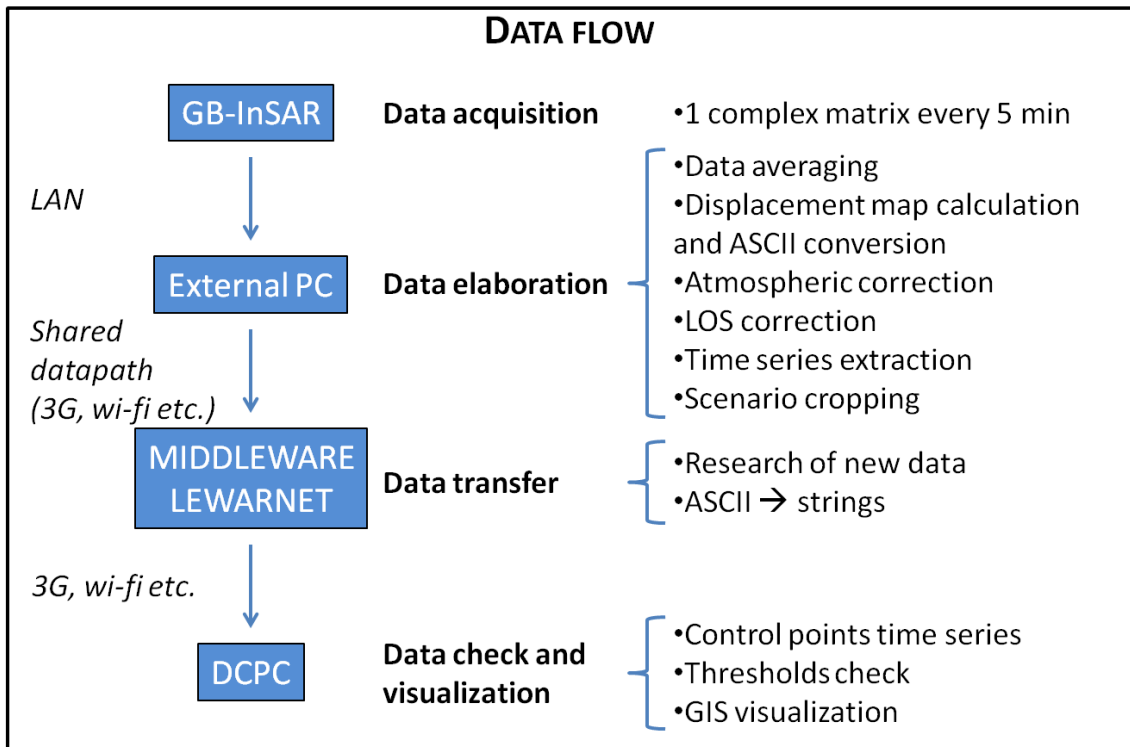
488
489 **Figure 4. Field of view of the slope from the GB-InSAR installation point.**

490 **5 Data management**

491 The most relevant matter of this monitoring was not as much related to the detection of landslide
492 movements threatening the highway, as to how a long-term monitoring performed with an
493 instrument providing huge amounts of data could have been run without resorting to large hard
494 drives nor to fast internet connections. ~~For this reason an appropriate data management (Figure 6)~~
495 ~~was developed.~~In fact, the monitoring area was covered by a 3G mobile telecommunication network
496 the installation was in an area where the only internet connection available was 3G, with a limit of 2
497 gigabyte data transfer per month and there was the need to reduce the massive data flow produced
498 by the radar.

499 For this reason, an appropriate data management (Figure 3) was developed and is here described.

500



501

502

Figure 3. Diagram showing the complete data flow from acquisition to final visualization.

503

5.1 Data acquisition

504

The GB-InSAR employed produced a single radar image, consisting in a 1001x1001 complex matrix, every 5 minutes. Each one is around 8 Megabytes large, resulting in more than 2 Gigabytes of data produced every day.

506

507

This amount of data represented an issue for both store capacity and data transmission.

508

5.2 Data elaboration

509

After being acquired, data were then transferred through LAN connection to the external PC implementing a dedicated Matlab script locally performing the actions described as follows.

510

511

5.2.1 Data averaging

512

In order to reduce the noise normally affecting radar data (especially in vegetated areas), the images acquired every 5 minutes were also averaged using all data of the previous 8 and 24 hours. Then images averaged on 24 hours have been used to calculate daily displacement maps, every 8 hours to create 8h displacement maps and non-averaged images to calculate 5 minutes displacement maps. These time frames have been selected based on the characteristics of the slope movements and signal/noise ratio in the investigated area.

517

518

Averaging is also a mean to make a good use of a high data frequency, since it enables to reduce the memory occupied in the database as an alternative to their direct elimination.

519

520 5.2.2 Displacement map calculation and ASCII conversion

521 Each radar image can be represented as in Eq.1:

522
$$S_n = A_n \exp(j\varphi_n) \tag{1}$$

523 where A_n is the amplitude of the n^{th} image, φ_n its phase and $j = (-1)^{1/2}$ is the imaginary unit. The
524 displacement Δr occurred in the time period between the acquisition of S_1 and S_2 has been
525 calculated with the following (Eq.2):

526
$$\Delta r = (\lambda/4\pi) \cdot \Delta\varphi \tag{2}$$

527 where λ is the wavelength of the signal and

528
$$\Delta\varphi = \varphi_1 - \varphi_2 \tag{3}$$

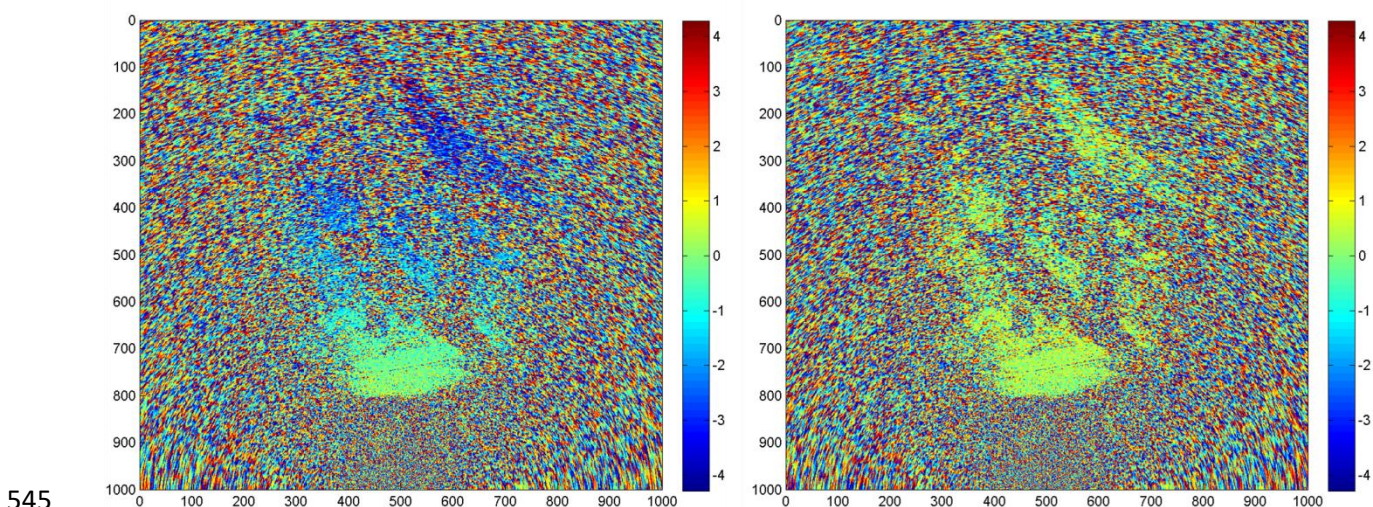
529 can be derived from:

530
$$S_3 = S_1 S_2^* = A_1 A_2 \exp[j(\varphi_1 - \varphi_2)] \tag{4}$$

531 As a result, an ASCII file, only containing the information relative to the displacement for each
532 pixel, was obtained.

533 5.2.3 Atmospheric correction

534 One of the major advantages of GB-InSAR is the capability to achieve sub-millimeter precision.
535 However, this can be severely hampered by the variations of air temperature and humidity,
536 especially when long distances are involved. Usually, atmospheric correction is performed by
537 choosing one area considered stable, taking into account that every displacement value different
538 from 0 is due to atmospheric noise and assuming that this offset is a linear function of the distance.
539 Based on this relation all the displacement map is corrected. In our case the whole scenario has been
540 selected and then only the potential unstable zones and those with a weak or incoherent
541 backscattered signal were removed. The remaining areas were then considered stable and therefore
542 were used for calculating the atmospheric effects. This results in a larger correction region that
543 enables a statistical correlation between the atmospheric effects and the distance and therefore the
544 calculation of a site-specific regression function that may not necessarily be linear (Figure 4).



546 **Figure 4. The color bar is expressed in mm; green indicates stable pixels, while blue and red**
547 **respectively movement toward and away from the GB-InSAR. Left: raw interferogram showing**
548 **artificial displacement increasing linearly with distance (as typical of atmospheric noise). Right: the**
549 **same interferogram after the atmospheric correction.**

550 5.2.4 Line of sight correction

551 The availability to detect only the LOS (Line Of Sight) component of the displacement vector
552 represents one of the main limitations of the GB-InSAR technique. A method to partially overcome
553 this limitation has been applied in this paper, following the procedure described in Colesanti &
554 Wasowski, 2006 and later in Bardi et al. 2014 and 2016. [Other methods have been employed by](#)
555 [Cascini et al. \(2010; 2013\).](#)

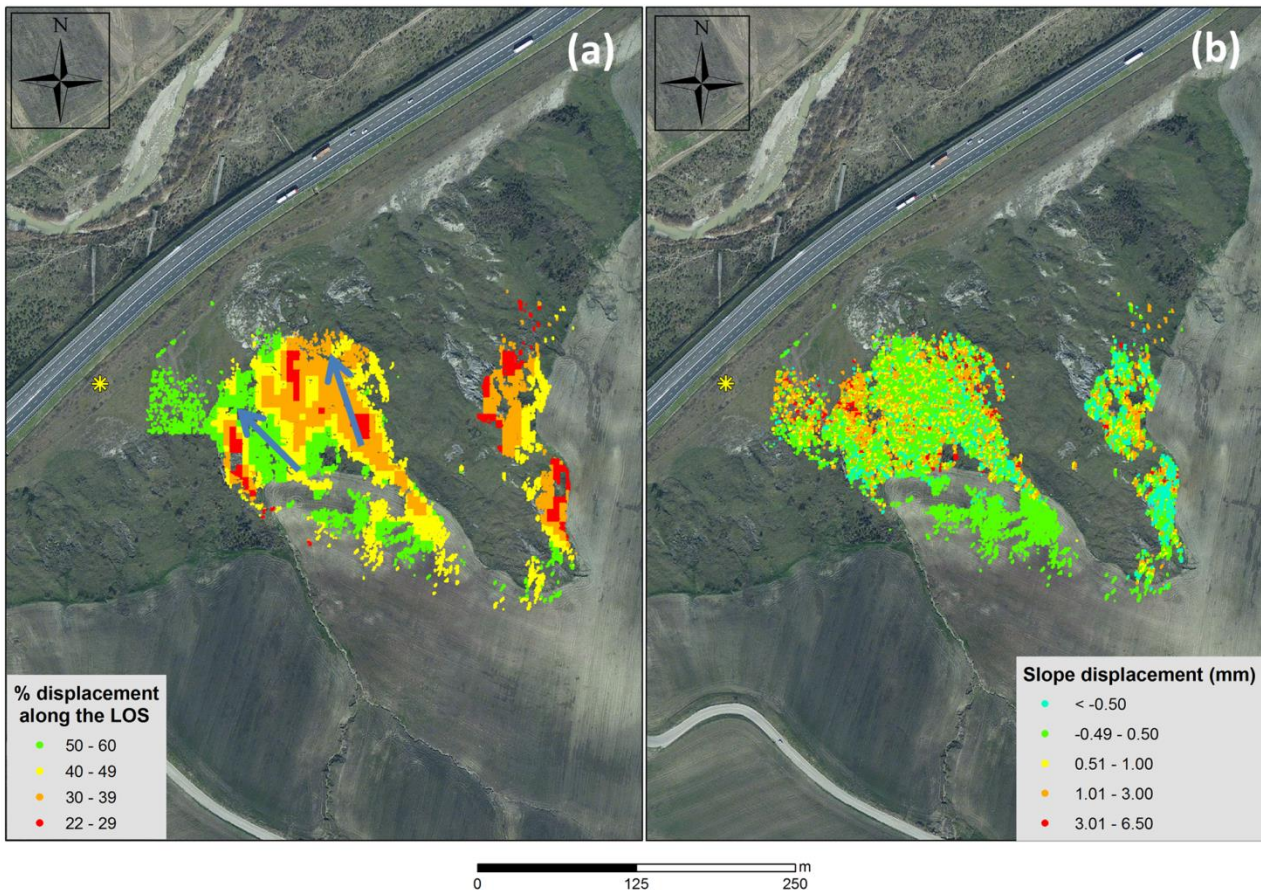
556 Assuming the downslope direction as the most probable displacement path, radar data have been
557 projected on this direction. Input data as the angular values of *Aspect* and *Slope* have been derived
558 from the *Digital Terrain Model* (DTM) of the investigated area; furthermore, azimuth angle and
559 incidence angle of the radar LOS have been obtained.

560 After calculating the direction cosines of LOS and Slope (respectively functions of azimuth and
561 incidence angles and aspect and slope angles) in the directions of Zenith (Z_{los} , Z_{slope}), North (N_{los} ,
562 N_{slope}) and East (E_{los} , E_{slope}), the coefficient C is defined as follow (Eq. 5):

$$563 \quad C = Z_{los} \times Z_{slope} + N_{los} \times N_{slope} + E_{los} \times E_{slope} \quad (5)$$

564 C represents the percentage of real displacement detected by the radar sensor (Figure 5A).

565 The real displacement (D_{real}) is defined as the ratio between the displacement recorded along the
566 LOS (D_{los}) and the C value (Figure 5 B).



567

568 **Figure 5. (a) C values map. Blue arrows indicate the downslope direction. (b) Cumulated displacement**
 569 **values projected along the downslope direction, referred to a period between 1 July 2014 and 1**
 570 **November 2014. The yellow asterisk in the left of the images represents the location of the GB-InSAR.**

571 Assuming that the studied landslide actually moves along the downslope direction, the GB-InSAR
 572 detectable real displacement percentage ranges between 22 and 60 % (Figure 5A).

573 In Figure 5B, an example of slope displacement map has been shown. Here, cumulated
 574 displacement data related to a period between 1 July and 1 November 2014 have been projected
 575 along the downslope direction. Data show as the area can be considered stable in the referred
 576 period; maximum displacement values of 4 mm in 4 months (eastern portion of observed scenario)
 577 can be still considered in the range of stability.

578 5.2.5 Time series extraction

579 In order to allow for a fast data transfer and velocity threshold comparison, some representative
 580 control points were selected, aimed at providing cumulated displacement time series. Control points
 581 were retrieved from the same displacement maps calculated as described in paragraph 5.2.2 and
 582 therefore can be relative to a time frame of 5 minutes, 8 hours or 24 hours.

583 In case of noisy data, instead of having a time series relative to a single pixel, these can be retrieved
 584 from a spatial average obtained from a small area consisting of few pixels.

585 5.2.6 Scenario cropping

586 Typically, the field of view of a GB-InSAR is larger than the actual area to be monitored. In fact, a
587 portion of the radar image may be relative to the ground, sky, areas geometrically shadowed or
588 covered by dense vegetation. These may be of no interest or even containing no information at all.
589 For the case here studied around 50% of a radar image had a low coherence and was for all practical
590 purposes, unusable. Therefore, a cropping of the ASCII displacement map occurred in order to
591 frame only the relevant area.

592 5.3 *Data transfer and visualization*

593 The interferometric data generated by GB-InSAR, after the pre-processing and proper correction
594 previously described, are ready for transfer to the DCPC. The transmission of these data to the
595 DCPC is mediated by the middleware, which interrogates the GB-InSAR for tracking the state,
596 detects the newest data, and reorders and marks them to properly build data time series to be
597 transferred to DCPC.

598 Subsequently, the middleware manages communications with the DCPC, according to the
599 implemented ad-hoc protocol. This ensures the security of data providers through encrypted
600 authentication mechanisms, it allows for recovering missing or partially transmitted data, thus
601 avoiding information loss, and provides data acquired by the sensors to the DCPC in a standardized
602 format, JSON, able to guarantee uniformity between the various information provided by the
603 various sensors types. All these particular features fully justify the adoption of an ad-hoc protocol
604 for data transfer, instead of using a standard protocol such as FTP.

605 The data files produced by the GB-InSAR have already been locally pre-processed and result in a
606 matrix expressed in ASCII code; the dimensions of the matrix are known and range from 1x1 (for
607 the displacement of single control points) to 1001x1001 (for uncropped displacement maps). Before
608 encapsulating these data in the message to be transferred to DCPC, the middleware converts them
609 from ASCII code to character strings, using the standard coding ISO / IEC 8859-1, so being able to
610 obtain a data compression with a factor equal to ≈ 8 .

611 Eventually the DCPC is entrusted for cumulating the displacements relative to the control points,
612 which are compared with the respective thresholds, and for visualizing the displacement maps as
613 WebGIS layers, thus enabling data validation and the evaluation of the extension of moving surface.

614 **6 Early warning procedures discussions**

615 The GB-InSAR is part of a larger early warning system (LEWIS) which also includes other
616 monitoring systems and simulation models. Therefore, to understand how GB-InSAR data can be
617 used in an early warning perspective, it is necessary to make reference to LEWIS as a whole.

618 Any information, coming from the investigated sites and subsequently processed also by using the
619 simulation models, is used to define an intervention model. This is based on the following elements:
620 event scenarios, risk scenarios, levels of criticality, levels of alert.

621 Event scenarios describe the properties of expected phenomena in terms of dimension, velocity,
622 involved material and occurrence probability. Occurrence probability depends on the associated
623 time horizon, which should be equal to few hours at most, in the case of early warning systems.

624 Evaluation of occurrence probability is carried out by using information from monitoring systems
625 and/or from outputs of adopted mathematical models for nowcasting. All the properties, to be
626 analyzed for event scenarios, are listed below; a subdivision in classes is adopted for each one:

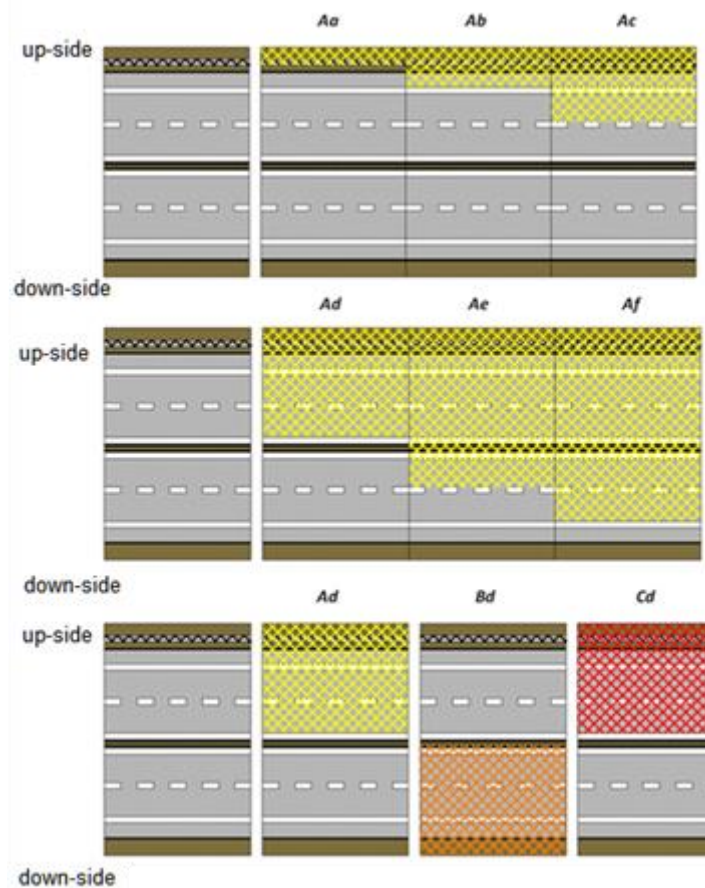
- 627 • landslide velocity (5 classes from slow to extremely rapid);
- 628 • landslide surface (5 classes from very small to very large);
- 629 • landslide scarp (5 classes from very small to very large);
- 630 • landslide volume (5 classes from extremely small to large);
- 631 • thickness (5 classes from very shallow to very deep);
- 632 • magnitude (3 classes: low, moderate, high), which combines the previous information;
- 633 • involved material (mud, debris, earth, rock, mixture of components);
- 634 • occurrence probability (zero, low, moderate, high, very high, equal to 1).

635 While some of the aforementioned parameters are determined by geological surveys, landslide
636 velocity is directly derived from monitoring data (such as those collected by GB-InSAR). Landslide
637 surface can be determined by geomorphological observation but is precisely quantified by GB-
638 InSAR, thanks to its capability of producing 2D displacement maps.

639 Risk scenarios can be firstly grouped in the following three classes:

- 640 A. mud and/or debris movements which could induce a friction reduction [between the vehicles](#)
641 [and the tar](#) and [therefore](#) facilitate slips;
- 642 B. road subsidence induced by landslides that could drag or drop vehicles;
- 643 C. falls of significant volumes and/or boulders that could crush or cover vehicles and constitute
644 an obstacle for others vehicles.

645 For each previous risk scenario, six sub-scenarios can be identified ~~on the basis of~~[based on](#) the
646 number of potentially involved infrastructures, carriageways and lanes (a. hydraulic infrastructures
647 and/or barriers, b. only emergency lane, c. lane, d. fast lane, e. fast lane of the opposite carriageway,
648 f. lane of the opposite carriageway). Thus, all possible risk scenarios are 18 (Figure 6)-, indicated
649 with a couple of letters (Capital and small).



650
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Figure 6. Top and middle: possible risk scenarios involving the scenario A (landslides that could reduce friction) to increasing sectors of the highway. Bottom: combinations of scenarios with different several types of phenomena (A, B, C) affect the emergency lane, lane and fast lane.

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The following information is provided to DCPC:

- Measurements from sensors
 - Model outputs
- and four states are identified for each of them:
- state 0 = no variation
 - state 1 = small variation
 - state 2 = moderate variation
 - state 3 = high variation.

662
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In practice, for the GB-InSAR, such states are delimited by fixed velocity values (thresholds). In this application values have been selected according to the gathered data, the first threshold being just above the instrumental noise; the remaining have been set based on expert judgement waiting for a more robust calibration, which is possible only after at least a partial mobilization of the slope. Anyway, the system is open to any method for determining thresholds (Crosta and Agliardi, 2003; Du et al., 2013; Carlà et al., 2016a) and also to the use of other parameters (acceleration for example).

669
670

Besides information from sensors and models, other information is obtained from meteorological and hydrological models (named as indicators).

671 Indicators comprise weather forecasting and output of FLAIR and Sushi models (Sirangelo et al.
 672 2003; Capparelli and Versace 2011) on the basis of observed and predicted (for the successive six
 673 hours) rainfall heights.

674 Two states are defined for indicators:

- 675 • state 0 = no variation or not significant,
- 676 • state 1 = significant variation.

677 To sum up, DCPC has the following information in any moment:

- 678 ▶ state (0, 1) of indicators (IND),
 - 679 ▶ state (0, 1, 2, 3) of sensors and models running for the specific highway section (SEN),
- 680 and, on the basis of these states, four different decisions can be made by DCPC, one of which with
 681 three options.

682 All the possible decisions are illustrated in Table 1, in which the weight of the several sensors is
 683 assumed to be the same. Based on the notices of criticality levels provided by the DCPC, and on its
 684 own independent evaluations, the CCC issues the appropriate warning notices (Surveillance, Alert,
 685 Alarm and Warning) and makes decisions about the consequent actions.

686

State of sensors and/or models	DCPC decisions
All INDs and SENs are S0	0 - no decision
At least one IND is S1 and all SENs are S0	1 – SOD (Sensor On Demand) activation
At least one SEN is S1	2 – to intensify the presence up to 24 hours/day
At least n SENs are S1 or at least one SEN is S2	3/1 – to issue a notice of ordinary criticality (level 1)
At least n SENs are S2 or at least one SEN is S3	3/2 - to issue a notice of moderate criticality (level 2)
At least n SENs are S3	3/3 - to issue a notice of high or severe criticality (level 3)

687

Table 1. DCPC possible decisions.

688 The information of each sensor and the results produced by the models are used to assess, in each
 689 instant, the occurrence probability of an event scenario in the monitored areas and the possible risk
 690 scenarios.

691 This combination of heterogeneous data was carried out by identifying for each sensor and model a
 692 typical information (displacement, precipitation, inclination, etc.), evaluating the state in each
 693 instant, according to a threshold system, and combining this result for all sensors placed in a
 694 monitored geomorphological area.

695 The ~~final~~ result is constituted by the occurrence probability of an event scenario, that is associated
 696 with a specific action by the DCPC. In particular, if the occurrence probability is low, moderate or
 697 high it is necessary to issue a notice of criticality (ordinary - Level 1, moderate - Level 2, High -
 698 Level 3) to the CCC.

699 The DCPC sends two types of information:

- 700 1) criticality state of the single monitored geomorphological unit,
- 701 2) criticality state of the whole area.

702 The adopted communication protocol between the two centers for the exchange of information was
703 carried out through a web service provided by the CCC, using the classes and attributes of the
704 methodology named Datex II (which is a protocol for the exchange of traffic data). The use of the
705 web service allowed to ensure the interoperability of data between the two centers, regardless of the
706 used hardware and software architecture, through a persistent service capable of ensuring an
707 immediate restoration of the connections, in case of malfunction and a continuous monitoring
708 between the two centers, even in the absence of criticality.
709

710 **7 Conclusions**

711 The GB-InSAR is a monitoring tool that is becoming more and more used in landslide monitoring
712 and early warning, especially thanks to its capability of producing real-time, 2D displacement maps.
713 On the other hand, it still suffers from some drawbacks, such as the limitation of measuring only the
714 LOS component of a target's movement and logistic issues like those owing to a massive
715 production of data that may cause trouble for both storing capacity and data transfer. In particular,
716 the latter is a more and more common problem of advanced technologies that are able to produce
717 high quality data with a high acquisition frequency, which may leave the problem of find the
718 balancing between exploiting all the information and at the same time avoiding unnecessary
719 redundancy.

720 These problems have been addressed when a GB-InSAR was integrated within a complex early
721 warning system (LEWIS) and only a limited internet connection was available. This situation
722 required that a series of pre-elaboration processes and data management procedures took place in
723 situ, in order to produce standardized and reduced files, carrying only the information needed when
724 it was needed. The procedures mainly concerned the transmission of data averaged over determined
725 time frames, proportionate with the kinematics of the monitored phenomenon. ~~Before~~Previously,
726 transmission data were also corrected (both in terms of atmospheric noise and LOS) and reduced,
727 by filtering out the information relative to the amplitude of the targets, by eliminating the areas not
728 relevant for the monitoring and by transforming the matrices into strings.

729 As a result, GB-InSAR data converged into the early warning system and contributed to it by
730 producing displacement time series of representative control points to be compared with fixed
731 thresholds. Displacement maps were also available for data validation by expert operators and for
732 retrieving information relative to the surface of the moving areas.

733

734 *Competing interests.* The authors declare that they have no conflict of interest.

735

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744 along the A16 highway.

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