

1 **HYDROLOGICAL DEFORMATION OF KARST AQUIFERS DETECTED BY GPS MEASUREMENTS,**
2 **MATESE MASSIF, ITALY**

3 **Authors:** ¹Leone Guido*, ²D'Agostino Nicola, ¹Esposito Libera, ¹Fiorillo Francesco

4 **Affiliations:** ¹Department of Sciences and Technologies, University of Sannio, Benevento, Italy; ²Istituto Nazionale
5 Geofisica e Vulcanologia, Rome, Italy

6 **Corresponding author:** Leone Guido*, email: guleone@outlook.com

7 **Abstract**

8 We used hydrological and geodetic observations to characterize the effects of hydrological forcing on the displacements
9 observed by GPS (Global Positioning System) measurements collected by stations located in the karst area of the Matese
10 massif (Apennines, central-southern Italy). The latter is one of the main karst massifs of the central-southern Apennines,
11 characterized by steep slopes, high mountain peaks (up to 2050 m a.s.l.), and wide endorheic areas, playing a fundamental
12 role in recharging the large groundwater resource of the massif. Integrated statistical analysis of rainfall and spring
13 discharge time series provided insight into timescales characterizing the hydrological dynamic of the Matese massif,
14 highlighting how hydraulic conditions of the water table at a specific time mainly depends on long antecedent periods of
15 rainfall. Nevertheless, intense daily rainfall occurring during the wet season are responsible of the abrupt increase of the
16 discharge of the Matese's springs, which show a typical karst behavior. Statistical analyses of time series show robust
17 correlations between hydrological conditions of the karst aquifer and GPS displacements observed at stations placed over
18 and around the massif. We find that the observed outward and inward deformations of the massif (horizontal dilatation
19 and contraction) are controlled by water table variations, which are in turn controlled by temporal variations of the
20 groundwater recharge due to rainfall. The detected patterns of deformation are intimately related to the seasonal and
21 multi-year characteristics of the recharge/discharge processes, and allowed us to track with geodetic measurements the
22 different phases of the water cycle in the karst aquifer of the Matese massif.

23 **Key words:** *Karst aquifer, deformation, GPS, spring discharge, Matese massif*

24 **INTRODUCTION**

25 The recent development of GNSS/GPS (Global Navigation Satellite System/Global Positioning System) networks
26 (Blewitt et al. 2018) has shown that space-time redistribution of surface and groundwater masses determines a significant,
27 measurable deformation of the Earth surface, providing an additional tool ("*hydrogeodesy*") for studying the water
28 circulation (White et al. 2022). In the last years, GPS (Global Positioning System) continuous monitoring has shown that
29 recharge/discharge phases of the karst aquifers hosted in the Mesozoic carbonates of central-southern Italy are strongly
30 correlated with measured ground displacements (Silverii et al. 2016; Silverii et al. 2019), and modulate the continuous,
31 long-term (~3-4 mm/yr) tectonic extension across the Apennines (D'Agostino 2014). Two main temporal patterns have
32 been highlighted: a seasonal oscillation controlled by seasonal fluctuations of the aquifer hydrological stage, and a multi-
33 year signal (long-term fluctuations) controlled by the occurrence of isolated or subsequent wet and dry periods (Silverii
34 et al. 2016; Silverii et al. 2019; D'Agostino et al. 2018). In particular, an increase in frequency of meteorological dry (or
35 drought) periods, along with an increase in mean annual temperature, have been observed in central-southern Apennine
36 over the last thirty years, and appears the main cause of the decline of the groundwater storage observed for karst massifs
37 of Mediterranean area (Fiorillo et al. 2021; Leone et al. 2021).

38 The strongest hydrological signal has been observed in horizontal components of displacements measured close to or
39 inside karst aquifer areas, and has been associated to variations of the groundwater storage, though the deformation signal
40 extended beyond the outcropping surface of the carbonate karst rocks (Silverii et al. 2019; Devoti et al. 2018).
41 As an example of these observations, **Figure 1** shows the horizontal daily displacements of a GPS station (*VAGA*) placed
42 inside the area of the Matese karst massif (**Figure 2**), in southern Italy, together with daily discharge of a basal karst
43 spring (Maretto spring, located 13 km from *VAGA*) draining the saturated zone of the karst aquifer and located along its
44 southern side. The temporal correlation between hydrological data/trend and measured ground displacement is evident.
45 In some cases, two displacement peaks can be observed in the wet season, both following peaks in the spring discharge.
46 As the latter reflects the water table stage in the aquifer (Fiorillo 2011), the simplest explanation argues for a control of
47 the hydraulic pressure in the aquifer saturated zone on the measured displacements.
48 The present study was built upon this observation and aims to describe the crustal deformation detected for the Matese
49 karst massif, where several GPS stations are located inside the karst area, and others along its boundary. The observed
50 phenomenon has been framed inside the local climate and hydrogeological context, and specific hydrological and
51 statistical analyses were carried out to highlight the relationships between the different physical variables (rainfall, spring
52 discharge, and ground displacement) and the respective time series.

53 1. STUDY AREA

54 The Matese massif is located in the median sector of the Apennine mountain range, and has an area of 540 km² (**Figure**
55 **2**). The morphology of Matese massif is characterized by steep slopes and surface elevations between 50 and 2050 m
56 a.s.l. (the highest peak is the Miletto Mt.). More than 50% of the massif lies above 900 m a.s.l. and endorheic areas
57 (**Figure 2**) occupies the most elevated zones (Leone et al. 2022). A highly permeable limestone and limestone-dolomite
58 sequence of Triassic-Miocene age, with a thickness of 2500-3000 m, is exposed at the Matese massif. Faults have a
59 prevalent NW-SE and NE-SW trends and dissect the sequence. The massif is tectonically juxtaposed by thrust faults to
60 low permeability argillaceous complexes (Paleocene) and flysch sequences (Miocene) along the northern and eastern
61 sectors (**Figure 2**). Normal faults bound the massif along the southern and western sectors and divide the massif from the
62 Volturno alluvial Plain (Boncio et al. 2016, 2022; Galli et al. 2017). More specific insight into the geology of the area can
63 be found in the Geological Map of Italy of Istituto Superiore per la Protezione e la Ricerca Ambientale (ISPRA;
64 <http://sgi.isprambiente.it/geologia100k/>). The climate is Mediterranean: the maximum monthly precipitation occurs in
65 November and the minimum in July, while the temperature and potential evapotranspiration patterns are almost opposite
66 to that of precipitation, with the maximum occurring in July-August and the minimum in December-February. Because
67 precipitation primarily occurs during the non-warm season, its distribution allows for the highest recharge into the aquifer
68 (Fiorillo and Pagnozzi 2015). In the high-elevation zones (above 1000 m a.s.l.), precipitation is generally snowy during
69 winter.

70 1.1. Hydrological karst features

71 The Matese massif represents a huge karst aquifer feeding numerous small springs as well as large basal springs. The
72 basal karst springs are fed by the saturated zone of the aquifer, which is recharged by percolation thorough the vadose
73 zone, locally characterized by a thickness up to hundreds or thousands of meters. Recharge occurs via both diffuse and
74 concentrate way thanks to the presence of ponors and dolines located at the bottom of endorheic areas (Leone et al. 2022).
75 The main spring groups are located along the southern side of the karst massif, inside the Piedimonte Matese and Telesse
76 Terme villages, and along the northern side, inside the Bojano village. These basal karst springs are thus located in the
77 lowest ground zones, along the geological contact between karst and no-karst terrains.

78 The Torano spring (200 m a.s.l.) is located on the eastern cliff of a karst canyon and has a mean annual discharge of 2.0
79 m³/s (Fiorillo et al. 2021), while the Maretto spring (170 m a.s.l.) is located at the foot of a limestone-dolomite slope in
80 the Piedimonte Matese urban area, and has a mean annual discharge of 1.0 m³/s (Fiorillo et al. 2021). Both the Torano
81 and Maretto springs were tapped during the 1960s by the Campano aqueduct, which supplies the metropolitan area of
82 Naples.

83 The Grassano-Telese springs (50-55 m a.s.l.) are placed along the southern side of Montepugliano relief, and are made
84 by large fresh and sulfurous thermal springs closely located, with a mean annual discharge of 4.5 and 0.2 m³/s,
85 respectively (Fiorillo et al. 2019). Moreover, the Montepugliano relief is interested by a notable concentration of collapse
86 sinkholes (22 sinkholes over an area of less than 1.5 km²), connected to ascendant phenomenon of karst spring waters
87 (Fiorillo et al. 2019) and the area surrounding this relief is the most important non-volcanic thermal area of Campania
88 region (Rufino et al. 2021).

89 Along the northern side of the Matese massif, there are another three major groups of springs, having an overall mean
90 annual discharge of 2.80 m³/s (Civita 1969). These springs are fed by the karst system of the north-central sector of Matese
91 massif and are located along debris deposits covering a fault separating limestone from flysch sequences (Petrella and
92 Celico 2009). Most of the springs were tapped by Biferno Aqueduct during the 1960s and supply the Molise region. A
93 branch of this aqueduct is joined with Campano Aqueduct through a tunnel in the eastern sector of Matese massif.

94 Many other karst springs are present on the Matese karst massif, and further details can be found in Fiorillo and Pagnozzi
95 (2015).

96 As in many karst areas of the world, the surface hydrology of the Matese massif is characterized by numerous endorheic
97 areas, which have been exploited during the last century to generate hydroelectric energy (Leone et al. 2022). The largest
98 ones are connected to structural and tectonic features (faults, grabens), which control the form and dimension of the
99 depressions, which are also known as poljes. Currently, some of them have seasonal lakes and are drained by one or more
100 ponors. These endorheic areas have a fundamental role in the recharge processes of the karst aquifer, as rainfall cannot
101 escape from the spring catchment through runoff. In particular, it has been estimated that only ≈30% of the rainfall, net
102 of evapotranspiration, escape from the massif as runoff. The other amount (≈70%) recharges the aquifer (Fiorillo and
103 Pagnozzi 2015). A detailed investigation of the Matese karst massif has allowed for mapping hundreds of these endorheic
104 areas, most of them ($N = 271$) smaller than 0.33 km². In total, the endoreic areas occupy 167 of 540 km² of the entire karst
105 massif (Leone et al. 2022). The largest endorheic area is that of Matese Lake polje (43.6 km²), located between 1000 and
106 2050 m a.s.l. It is the major karst morpho-structural feature of the massif, whose evolution has been controlled mainly by
107 WNW-ESE normal faults. The Matese Lake is a permanent water reservoir, sustained by fine sediments and pyroclastic
108 ash coming from volcanos located along the Campania coast (Vesuvius and Phlegraean Fields), but it was modified by
109 hydraulic works to avoid its water losses into ponors.

110 The endoreic areas of the Letino and Gallo Lakes are effectively dam reservoirs. The lakes were created by the interruption
111 of the Lete and Sava Rivers, respectively, but can be considered part of wide endoreic areas. In fact, before dam
112 construction, both rivers sunk into two ponors located downstream the dams.

113 The Campitello endorheic area is another important karst feature of the Matese massif. The area gets flooded during rapid
114 snow melting periods and a temporary lake forms. A ponor located at 1417 m a.s.l. in the central part of the Campitello
115 plain drains the lake water, which dries out in several days.

116
117

118 **2. DATA AND METHODS**

119 **Table 1** shows information about stations and the relative daily time series used in this study, which are illustrated in
120 **Figure 3**. The various hydrological and geodetic observations extend over a period of about 15 years, with different data
121 coverages.

122 Daily rainfall and temperature data have been provided by Multi-risk Functional Center of Civil Protection of Campania
123 Region (<http://centrofunzionale.regione.campania.it/#/pages/dashboard>). Even if there are several rain gauges
124 surrounding the Matese massif, analyses were primarily carried out using the daily measurements of the Alife rain gauge
125 and thermometer (*S1*, 117 m a.s.l.), because of its continuous records, without interferences connected with snowy
126 precipitation. Rainfall time series was used to define the input hydrological signal controlling spring discharge and,
127 consequently, the ground displacements recorded at GPS sites. Instead, the Letino rain gauge (*S2*, 1050 m a.s.l.) was used
128 in simulation of the karst aquifer recharge, given its suitable elevation.

129 Discharges of Torano and Maretto springs were provided by the Aqueduct company managed by the Regione Campania.
130 These discharge data were used because of the continuity of the records and quality of the measurements at daily scale.

131 Stations from various GNSS/GPS networks were considered, which provided continuous raw displacement data. In
132 particular, the *VAGA*, *LNGN* and *PTRJ* stations belong to the INGV RING GNSS/GPS network (<http://ring.gm.ingv.it>),
133 the *VINC* station belongs to the NETGEO GNSS/GPS network (<http://www.netgeo.it>), and the *ALIF* station to the
134 Campania Regional GNSS/GPS network (<http://gps.sit.regione.campania.it/indexmain.php>). GPS data are reduced using
135 the Jet Propulsion Laboratory (JPL) GIPSY-OASIS II software (ver. 6.3) in a Precise Point Positioning (Bertiger et al.
136 2010) mode, updating the results of D'Agostino et al. (2018), where the full processing is described in detail. Since the
137 primary focus of this work is the deformation of hydrological origin, we removed the secular, tectonic, linear trends (see
138 D'Agostino 2014 for their tectonic implications), estimated with a weighted least-squares approach. When seasonal
139 signals are taken into account, typical daily scatter of daily position time series are ~1 mm and ~4 mm for horizontal and
140 vertical components, respectively (Blewitt et al. 2018).

141 Time series were managed and analyzed by MATLAB (ver. 2019b) software. In particular, cross-correlation analysis was
142 carried out to evaluate statistical correlations between rainfall and spring discharge, and between hydrological variables
143 and GPS-measured displacements. Fast Fourier Transform-based (FFT-based) analysis and Gaussian-weighted moving
144 average (Gaussian smoothing) were successively used to highlight the seasonal and multi-year fluctuations in time series.

145 **Table 1** Hydrological and GPS time series used in the study. The elevation (in m a.s.l.), latitude and longitude (in decimal
 146 degree), and length of the available time series are summarized for each station and karst spring.

		Location			Time interval of daily data	
		Elevation	Latitude	Longitude		
GPS station	<i>Within karst massif</i>	VAGA (Valle Agricola)	736	41.415	14.234	15 Mar 2005-31 Oct 2020
		LNGN (Longano)	1021	41.501	14.253	2 Jun 2006-3 Dec 2016
		PTRJ (Pietraraja)	1065	41.364	14.529	19 Jul 2006-31 Oct 2020
	<i>Outside karst massif</i>	ALIF (Alife)	109	41.327	14.335	25 Sep 2008-31 Oct 2020
		VINC (Vinchiaturro)	514	41.468	14.562	28 Jan 2012-31 Oct 2020
		S1 (Alife)	117	41.339	14.334	15 Mar 2005-31 Oct 2020
Rain gauge and thermometer	S2 (Letino)	1050	41.454	14.253	31 Oct 2008-31 Oct 2020	
	Maretto	170	41.358	14.370	15 Mar 2005-31 Oct 2020	
Karst spring	Torano	200	41.365	14.381	15 Mar 2005-31 Oct 2020	

147 **2.1. Cross-correlation analysis and rainfall data pre-processing**

148 The cross-correlation function describes the similarity, or correlation, between a time series x and the lagged versions of
 149 another time series y as a function of the time lag (k -lag). The similarity between time series is expressed by the Pearson
 150 correlation coefficient, $r_{xy}(k)$, which varies between -1 (maximum negative correlation) and $+1$ (maximum positive
 151 correlation).

152 In particular, the linear cross-correlation function, $r_{xy}(k)$, is defined as (Larocque et al. 1998):

153
$$r_{xy}(k) = \frac{C_{xy}(k)}{\sigma_x \sigma_y}$$

154 where $C_{xy}(k)$ is the covariance between the time series of x and y variables computed at time lag k , while σ_x and σ_y are the
 155 standard deviations of the time series. Typically, when analyzing the link between the climate and a specific groundwater
 156 variable (spring discharge, or water level), time series of the independent variable x is the time series of rainfall, while
 157 the lagged time series of y is represented by spring discharge. Instead, when evaluating the correlation between
 158 hydrological and GPS observations, x represents rainfall or spring discharge and y is the displacement.

159 In our investigation, cross-correlation analysis was used to evaluate statistical similarities between time series of the
 160 different variables and the time scale characterizing the processes under investigation.

161 First, it must be remarked that daily rainfall cannot completely explain measured spring discharge value at specific time,
 162 which rather depends on the long antecedent period of rainfall (Fiorillo and Doglioni 2010). Therefore, cross-correlation
 163 analysis was primarily used to estimate the period of antecedent rainfall explaining the observed spring discharge. This
 164 analysis involved the pre-processing of the daily rainfall data, to construct multiple continuous time series of the
 165 cumulative rainfall. These time series were obtained by summing daily values in a time window of fixed width, which
 166 has then been moved one step forward through the whole time series. In this study, window width varies from 1 to 360
 167 days.

168 **2.2. Detection of seasonal and multi-year fluctuations of hydrological and GPS time series**

169 A hydrological time series is a sequence of measurements of a specific hydrological variable in chronological order.
 170 Measured values of hydrological variables, such as daily rainfall, spring discharge, and air temperature, show temporal
 171 variations due to different physical processes which, in turn, continuously evolve over time.

172 Variations observed in hydrological time series are ascribable to five components, which are the long-term trend in mean
173 (and/or variance), seasonality, cyclicity, jumps and randomness (Beaumont 1982; Caswell 1991). In the hypothesis that
174 hydrological processes affect the motion of the land surface, and thus displacement measurements, the hydrological
175 patterns observable in hydrological time series would be recorded by GPS stations and recognizable in the displacement
176 recorded over the time.

177 In practice, GPS measurements are affected by a number of factors and include random and systematic noises from
178 different sources, which could make the hydrologically induced displacement difficult to recognize and quantify (Blewitt
179 2015). To this scope, the Fast Fourier Transform-based (FFT-based) analysis and Gaussian-weighted moving average
180 (Gaussian smoothing) were used in this study, allowing us to emphasize the short- and long-term oscillations in the time
181 series. The time patterns characterizing the hydrological series was then compared to those identified in GPS time series
182 to highlight any similarities among the two kinds of signals.

183 In particular, the Gaussian smoothing, was used to outline trend component of time series, that is, the long-term variations
184 in spring discharge, rainfall and ground motion. The Gaussian smoothing belongs to moving average smoothing methods
185 (Fei and Jin 2016), which are classical approaches used in estimating time series trend (Caswell 1991; Hyndman and
186 Athanasopoul 2018). Gaussian smoothing was preferred to classical moving average method in this study because it is
187 more effective in reducing the short-term fluctuations. A 4-year time window was used for Gaussian smoothing, to
188 eliminate the effect of short-term fluctuation due to hydrological seasonality.

189 On the other hand, the FFT-based filtering was applied to emphasize the short-term cyclic (seasonal) oscillations of time
190 series superimposed on the trend component. In this case, this method was preferred instead of Gaussian filtering as it
191 provided more suitable results for time series characterized by significant data gaps and data dispersion. More precisely,
192 the FFT-based analysis was applied to decompose time series into a series of sinusoids of varying amplitude, phase, and
193 frequency (Kramer 1996; Fleming et al. 2002; Box et al. 2016). The main components making a significant contribution
194 to the time series were selected by analyzing the Fourier amplitude spectra, which describe how the amplitudes of a signal
195 are distributed among different frequencies (Kramer 1996). Finally, smoothed time series were reconstructed by summing
196 together the few components having the highest amplitudes.

197 3. RESULTS

198 Time series analyses carried out in this study aimed to ascertain the link between hydrological processes and ground
199 displacement detected through continuous GPS measurements in the area of the Matese karst massif.

200 3.1. Relationships between rainfall and karst spring discharge

201 Daily rainfall does not fully explain the spring discharge at specific time. On the other hand, rainfall cumulated over long
202 time periods can describe the dynamic of the karst springs and, thus, of the karst aquifer (Fiorillo and Doglioni 2010).
203 Based on this consideration, cross-correlation analysis was performed between daily rainfall and spring discharge time
204 series to evaluate the period of cumulate rainfall which better explains the groundwater dynamic of the Matese karst
205 massif.

206 **Table 2** and **Figure 4** show the results of cross-correlation analysis between rainfall cumulated over different periods
207 (from 1 to 360 days) and daily spring discharge. It can be seen how the Pearson correlation coefficient $r_{xy}(k)$ increases
208 with increasing k -lag (in days) for short rainfall periods, while it is maximum when reducing k -lag and increasing the
209 period of cumulative rainfall beyond three months. This indicates that the spring discharge is mainly controlled by rainfall
210 cumulated over long-time periods, at least up to 270 days. As shown by **Table 2**, the highest correlation coefficient and
211 the lowest k -lag were obtained for rainfall cumulated over periods of 180 and 270 days.

212 **Table 2** Results of the cross-correlation analysis between daily time series of rainfall cumulated over period of different
 213 length (periods from 1 day up to 360 days) and discharge of Torano and Maretto karst springs. The maximum value of
 214 the Pearson correlation coefficient, $r_{xy}(k)$, and the relative k -lag (in days) are shown in table.

Period of cumulate rainfall	Torano spring		Maretto spring	
	$r_{xy}(k)$	k -lag	$r_{xy}(k)$	k -lag
1 day	0.121	97	0.141	93
7 days	0.245	92	0.287	92
15 days	0.318	88	0.375	91
30 days	0.408	85	0.485	90
60 days	0.511	60	0.610	78
90 days	0.600	47	0.686	51
120 days	0.644	27	0.729	43
180 days	0.742	1	0.776	10
270 days	0.689	0	0.705	0
360 days	0.597	0	0.489	0

215 *3.2. Relationships between hydrological variables and ground displacement*

216 The results of the analysis described in the previous section highlighted the overall timescales characterizing the
 217 hydrological dynamic of the Matese massif, where spring discharge at a specific time primarily depends on the rainfall
 218 which has occurred during the antecedent 180-270 days.

219 Starting from this result, a further cross-correlation analysis was carried out between hydrological records (180- and 270-
 220 day rainfall and spring discharge) and raw GPS displacement data. Results of the analysis are summarized in **Figure 5**
 221 and **Table 3**, and provided insight into the correlation between hydrological processes and ground displacement, and the
 222 average response time of the ground motion to variations of the aquifer hydrological stage.

223 We observe that the displacements in the horizontal plain are linearly (positively or negatively) correlated with
 224 hydrological stages of the aquifer. Conversely, if excluding the *LNGN* GPS stations, the vertical displacement at single
 225 sites would not show any clear correlation with the hydrological variables ($|r_{xy}(k)| < 0.3$).

226 Horizontal displacements recorded at *VAGA* GPS station (**Figure 5a**) are strongly, negatively correlated with 6-month
 227 rainfall and spring discharge. This indicate that the station moves south-westward, i.e., outward from the massif, toward
 228 the alluvial plain of the Volturno River, during high groundwater stages. The time lag between the hydrological forcing
 229 and displacement varies from few days up to 21 days, and it is lower for the east component than for the north component.

230 Displacements of the *ALIF* station (**Figure 5b**) appear consistent with those of the *VAGA* station. In fact, displacement in
 231 the north direction is negatively correlated with Torano spring discharge and 6-month cumulative rainfall, while there
 232 would be no clear correlation of the hydrological variables with the east and vertical components of the displacement.

233 For the *LNGN* GPS station (**Figure 5c**), the hydrological stages mainly affect displacements in the north direction. The
 234 north component is strongly, positively correlated with rainfall and spring discharge, highlighting that the station moves
 235 toward north during aquifer recharge and hydrological wet periods. The coefficient $r_{xy}(k)$ is maximum for k -lags between
 236 0 and 6 days. On the contrary, a time lag of 134 days characterizes the cross-correlation function associated with the east
 237 component, which appears weakly, negatively correlated with hydrological variables.

238 The east component of displacement of *PTRJ* GPS station (**Figure 5e**) is characterized by quite high, positive values of
 239 the cross-correlation coefficient. Differently from the *VAGA* and *LNGN* stations, the *PTRJ* station move toward east

240 during high hydrological stages. This hydrologically induced displacement appears lagged of 90 days with respect to
 241 spring discharge and 31 days with respect to 9-month cumulative rainfall. No correlation would appear for the north
 242 component with hydrological variables.

243 Data records of the *VINC* GPS station (**Figure 5d**) also appears in line with those of the stations placed on the karst
 244 massif. Horizontal displacements in the north and east directions are positively correlated with spring discharge and 6-
 245 month rainfall. Thus, the station would move toward both east and north during hydrologically wet periods and toward
 246 south and west during dry periods. Displacements are lagged with respect to the hydrological forcing and time lags are
 247 significantly shorter for the east component than for the north component.

248 **Table 3** Results of the cross-correlation analysis between hydrological time series and GPS time series. Statistically
 249 significant values of the Pearson correlation coefficient, $r_{xy}(k)$, and the relative k -lag (in days) are shown in table. The 9-
 250 month cumulative rainfall was considered for the *PTRJ* GPS station and the 6-month cumulative rainfall for the other
 251 stations.

GPS station	Displacement component	Cumulative rainfall		Torano spring		Maretto spring	
		$r_{xy}(k)$	k -lag	$r_{xy}(k)$	k -lag	$r_{xy}(k)$	k -lag
VAGA	<i>Nord</i>	-0.887	21	-0.838	17	-0.878	10
	<i>East</i>	-0.872	13	-0.814	10	-0.852	3
	<i>Vertical</i>	-	-	-	-	-	-
ALIF	<i>Nord</i>	-0.484	96	-0.510	74	-0.437	68
	<i>East</i>	-	-	-	-	-	-
	<i>Vertical</i>	-	-	-	-	-	-
PTRJ	<i>Nord</i>	-	-	-	-	-	-
	<i>East</i>	0.533	31	0.472	88	0.372	89
	<i>Vertical</i>	-	-	-	-	-	-
VINC	<i>Nord</i>	0.412	101	0.407	98	0.449	82
	<i>East</i>	0.452	41	0.459	9	0.317	9
	<i>Vertical</i>	-	-	-	-	-	-
LNGN	<i>Nord</i>	0.732	6	0.731	2	0.706	0
	<i>East</i>	-	-	-0.368	133	-0.359	130
	<i>Vertical</i>	0.388	0	0.348	0	0.357	0

252 In summary we observe a strong correlation between the horizontal displacements and the hydrological phases of
 253 recharge/discharge of the Matese massif. The time lags of cross-correlation values $|r_{xy}(k)| > 0.5$ between spring discharge
 254 and GPS displacements yield an average value of 16 days suggesting a very fast deformational response to the
 255 hydrological forcing. Conversely, the low values of cross-correlation coefficient associated with the vertical
 256 displacements imply their negligible contribution to the accommodation of the hydrological forcing.

257 3.3. Hydrological patterns in time series of the ground displacement

258 FFT-based and Gaussian filtering were carried out to emphasize short- and long-term oscillations in time series with the
 259 aim of ascertain any recognizable hydrological patterns in GPS measurements of the ground motion.

260 **Figure 3** compares time series of hydrological and GPS data. In particular, for each panel in Figure 3, the raw
 261 measurements, the long-term mean of the time series (\square ; black dashed line), the filtered time series computed from FFT-
 262 based analysis (continuous solid line), and the Gaussian smoothing (red dashed line) are plotted.

263 It should be noted that the mean displacement values in **Figure 3e-n**, $D_{\text{north}} = 0$ mm and $D_{\text{east}} = 0$ mm, represent the
 264 intersection of the north and east axes. Thus, positive displacement values indicate that a specific GPS station moves to
 265 the north and east quarters, while negative values indicate that the station moves towards the south and west quarters.

266 **Figures 3a, b** and **d** confirm results of cross-correlation analysis, as both short- and long-term variations of the rainfall
 267 cumulated over a period of 180 days (R_{180} , **Figure 3d**) closely follow spring discharge fluctuations (**Figure 3a, b**),
 268 highlighting that the hydrological stage of the massif at any time strongly depends on the previous period of rainfall. The
 269 range between $\square - \sigma$ and $\square + \sigma$ has been drawn in **Figure 3d**, which represents the normal rainfall condition in the
 270 reference period (\square and σ are the long-term mean and standard deviation of the time series, respectively). Drought years
 271 characterized by below-normal rainfall amounts can be observed in 2007, 2012, 2017. Drought years mark the end of
 272 decreasing rainfall periods, which are the main cause of groundwater storage drying and, thus, of spring discharge decline
 273 (**Figure 3a, b**). The recovery of the spring discharge following meteorological droughts is related to significant increases
 274 of the rainfall amounts, which are marked by peaks in the rainfall time series higher than $\square + \sigma$.

275 A seasonal component is present in time series of the ground displacement. In addition, the hydrological patter
 276 characterizing spring discharge and rainfall is well recognizable in the GPS time series of the horizontal displacement
 277 (**Figure 5e-n**), indicating that the karst massif responds to seasonal and long-term variations of the groundwater storage,
 278 as well as to abnormal increases or decreases of the recharge. Sequences of consecutive dry (or wet) years can be observed
 279 in **Figure 3d**, which give rise to periods of below average (or above average) groundwater storage and spring discharge.
 280 Such a low-frequency hydrological variation (red dashed line in **Figure 3**) affects the ground displacement and is clear in
 281 the long-time series of the *VAGA* GPS station (**Figure 3e, f**) and *PTRJ* (**Figure 3j**), and in shorter time series of *VINC*
 282 (**Figure 3i**), and *LNGN* (**Figure 3m**) stations as well. Furthermore, a strong hydrological signal is associated with
 283 droughts, such as those of 2007, 2012 and 2017, which are marked by low downward peaks in GPS time series.

284 Although a seasonal component was detected also in time series of the vertical displacement, the hydrological patterns
 285 are not clearly evident in this case. As already discussed by Silverii et al. (2019), a low signal-to-noise ratio generally
 286 characterizes time series of the vertical displacement of an individual GPS station, which could obscure the hydrological
 287 patterns, and data from several stations are required to enhance the signal-to-noise ratio.

288 3.4. Strain analysis

289 To investigate the overall deformation response of the Matese karst massif to hydrological forcing we calculated the
 290 temporal variation of the strain tensor in the area defined by the polygon obtained by joining the five GPS stations (**Figure**
 291 **6** and Supplementary Material). The spatial derivatives of the stations displacements u in the x (North) and y (East)
 292 directions define the components of the displacement gradient tensor:

$$293 \frac{\partial u_x}{\partial x}, \frac{\partial u_x}{\partial y}, \frac{\partial u_y}{\partial x}, \frac{\partial u_y}{\partial y}$$

294 The displacement gradients (extension positive) are then used to calculate the components of the symmetric strain tensor:

$$295 \varepsilon_{xx} = \frac{\partial u_x}{\partial x}, \quad \varepsilon_{yy} = \frac{\partial u_y}{\partial y}, \quad \varepsilon_{xy} = \varepsilon_{yx} = \frac{1}{2} \left[\frac{\partial u_x}{\partial y} + \frac{\partial u_y}{\partial x} \right]$$

296 The principal axes of the strain tensor are then calculated as:

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$$\varepsilon_1, \varepsilon_2 = \frac{\varepsilon_{xx} + \varepsilon_{yy}}{2} \pm \sqrt{\left(\frac{\varepsilon_{xx} - \varepsilon_{yy}}{2}\right)^2 + \varepsilon_{xy}^2}$$

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where $\varepsilon_1, \varepsilon_2$ are the magnitudes of the most extensional (extension positive) and compressional eigenvalues of the strain rate tensor. The direction θ of the most extensional eigenvalue (counter-clockwise from East) is finally given by the following relationship:

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$$\tan(2\theta) = \frac{2\varepsilon_{xy}}{\varepsilon_{xx} - \varepsilon_{yy}}.$$

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To emphasize the seasonal and multi-year response we used the horizontal components of the FFT-filtered time series (**Figure 3**) of five stations in and around the Matese massif, resampled at time steps of 0.01 years. For each epoch in the time interval of simultaneous activity of the five stations (January 2012–December 2016), we estimated with a weighted least-squares approach the components of the displacement gradients and the strain tensor. **Figure 6** illustrates two snapshots of the continuous strain tensor, representative of dry (low water table, November 2012) and wet (high water table, May 2013) conditions (the complete continuous evolution is available as an animation in Supplementary Material). The inset in **Figure 6** shows the evolution of the dilatational strain ($\varepsilon_{xx} + \varepsilon_{yy}$) compared to the discharge of the Torano and Maretto springs. A low water table condition (November 2012) corresponds to inward displacements and negative (contractional) dilatational strain up to -4×10^{-7} relative to the average of the 2012–2017 period. A high water table condition (May 2013) corresponds to outward displacements of the stations in the Matese massif and positive (extensional) dilatational strain. The maximum amplitude of the average hydrological strain across the Matese massif is approximately 6×10^{-7} and is probably not homogeneously distributed, although the lack of a dense station coverage halts a more accurate analysis.

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4. DISCUSSION AND CONCLUSION

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This study deals with the link between karst aquifer hydrological stage and crustal deformation, detected for the Matese massif. Time series of hydrological data were used to represent the hydrological stage of the karst system, which is the hydraulic condition of the karst aquifer at a certain time. In turn, the hydraulic conditions depend on water table level in the saturated zone, that is the groundwater storage, as well as the water level in shafts connected to the saturated zone. Shaft has a crucial role in groundwater dynamics since they drain water rapidly, causing a fast recharge into the saturated zone and, eventually, the raising of the water table level (Fiorillo 2011). The abrupt increases of the discharge observed in spring hydrographs following intense rainfall events are also generally associated with the rise of the water level in conduits above the saturated zone (Bonacci and Zivaljevic 1993; Bonacci 1995; Drogue 1980; Ford and Williams 2007). Direct measurements of the water level into karst aquifer are unavailable for the Matese massif, and records of daily discharge of Torano and Maretto springs were considered as a proxy for the water table fluctuations. To investigate more in detail the relationship between spring discharge and GPS measured displacement, we simulated the recharge process during the 2009–2010 hydrological year (**Figure 7**), using the recharge model developed by Fiorillo et al. (2015a) and calibrated by Fiorillo and Pagnozzi (2015). This model transforms the daily rainfall into excess rainfall, P_{exc} , through the hydrological balance of the soil (**Figure 7b**). The P_{exc} is the amount of rainfall which is free to infiltrate in depth into the vadose zone (recharge) or to run off when a specific threshold (24.3 mm) is exceeded. This threshold value has been calibrated from long term annual hydrological balance of karst massif (Fiorillo et. al. 2015a; Fiorillo and Pagnozzi 2015). The runoff allows the surficial water to leave the karst massif without any contribute to recharge process. In the endorheic areas, however, surficial water can be considered as “internal runoff” (White 2002), that totally contributes to recharge by concentrate infiltration into ponors and dolines or leakage from bottom of endorheic areas. According to our model

335 rainfall is retained as soil moisture into the soil mantle up to the field capacity until the end of November or it is lost
336 through evapotranspiration processes (**Figure 7b**). During this period, the hydrograph highlights the recession of the
337 spring (**Figure 7c**). The rainfall of early December causes the beginning of the recharge processes and interrupts the
338 spring discharge recession. Snowy precipitations occurring during winter postpone the infiltration processes. Looking at
339 **Figure 6c**, GPS displacements also invert the trend in December. Successively, the intense rainfall of early January
340 generates the diffuse and concentrate recharge, which in turn causes the rapid increase of the spring discharge, and the
341 acceleration of the GPS station as well (**Figure 7c**). Up to mid-March, recharge processes sustain high flow rate, then
342 discharge decreases due to decrease of rainfall. The GPS site experiences the maximum displacement in this period, which
343 seems shifted forward of about 1 month with respect to the peak of the spring discharge. The end of recharge processes
344 starts from early June, and soil moisture begins to accumulate a water deficit. Spring discharge decreases since April, and
345 it can be considered no longer affected by recharge processes from June. GPS displacement also decreases in the same
346 period.

347 Our observations indicate a close relationship between hydrological conditions in the karst aquifer and the observed
348 deformation of the Matese massif. **Figure 8** shows two different stages of the water table (low- and high-stand),
349 corresponding to a low and high spring discharge condition in **Figure 6**, and the connected deformation observed at the
350 GPS stations.

351 Similar to other areas in the central and southern Apennines (Silverii et al. 2016; Silverii et al. 2019), the hydrological
352 forcing dominantly affects the horizontal deformations, with minor impact on vertical displacements. As observed by
353 Silverii et al. (2016) for the Irpinia region, the vertical component displays seasonal and multiyear signals more
354 homogeneously distributed in space, independently from extent of the karst aquifers and closely correlated with regional
355 estimates of total water storage. The different sensitivity of the horizontal components to the hydrologically induced
356 deformation processes, allow us to temporally resolve the maximum hydraulic head in the karst aquifer. We suggest that
357 the main mechanism controlling the observed horizontal transient signal is the effect of time-dependent water pressure
358 which induces the rock mass strain, void and fluid-filled volume change, as consequence of the poroelastic behavior of
359 the karst aquifer. Conductive water-filled fracture systems could have an important role in controlling the anisotropic
360 deformation along the different directions. However, the extent at depth of this process is unclear. In other part of the
361 Apennines (D'Agostino et al. 2018) hydrological seasonal modulation appear to control the nucleation of low-magnitude
362 seismicity in the shallow suggesting that the effect of hydrological forcing extend to the first kilometers (0-6 km) of the
363 crust.

364 Finally, we remark that the observed pattern of hydrologically-related deformation is intimately connected to the specific
365 pattern of recharge of the karst aquifers in the Apennines (Fiorillo et al. 2015b). Slow groundwater circulation and
366 multiyear variations of the water table elevation, allows significant variations of the stored groundwater, that, in turn,
367 control the observed pattern of deformation.

368 **Supplementary Material (SI)**

369 Movie 1. Animation of horizontal GPS displacements at stations in the Matese massifs (red vectors) and the associated
370 evolution of the principal axes of the strain tensor (diverging/converging green vectors) in the polygon joining GPS
371 stations (black line). Areas with outcropping carbonate rocks are emphasized with a blueish tone. The diagram in the
372 upper right shows the temporal evolution of the horizontal dilatational strain ($\epsilon_{xx} + \epsilon_{yy}$) and the discharge at Torano-
373 Maretto springs highlighting the strong correlation between horizontal deformation and hydraulic conditions in the Matese
374 aquifer.

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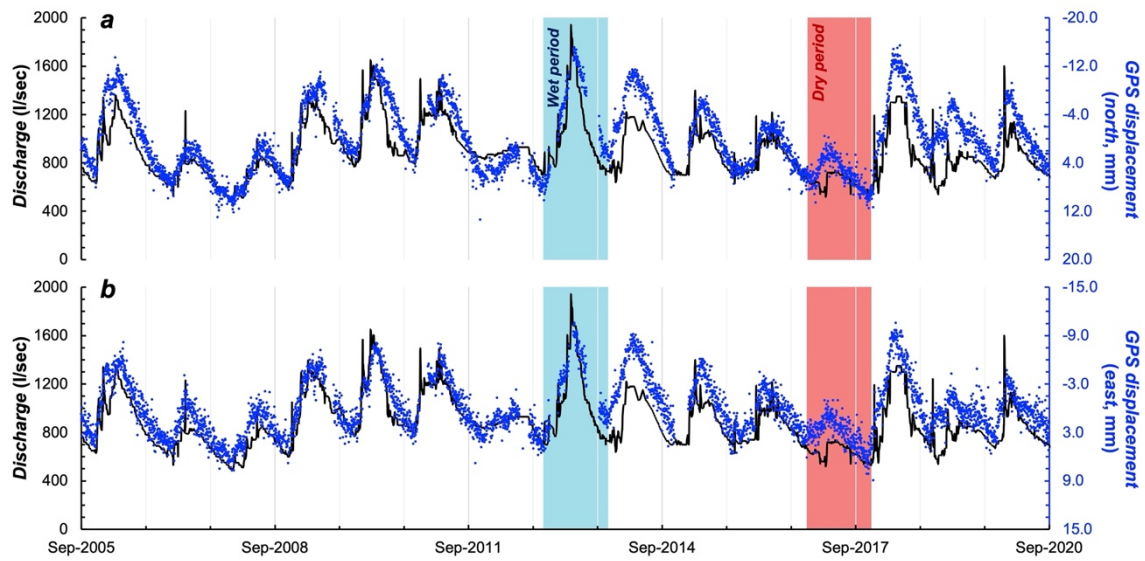
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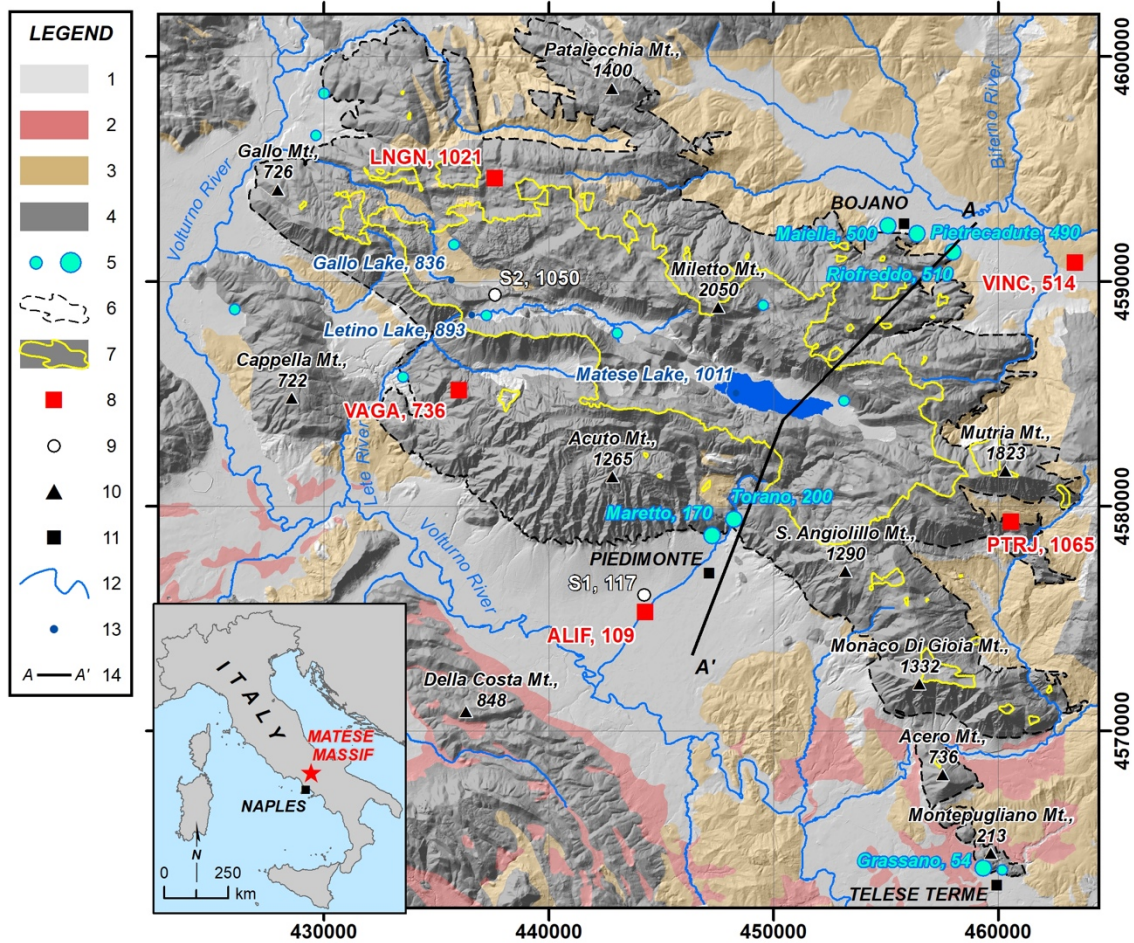
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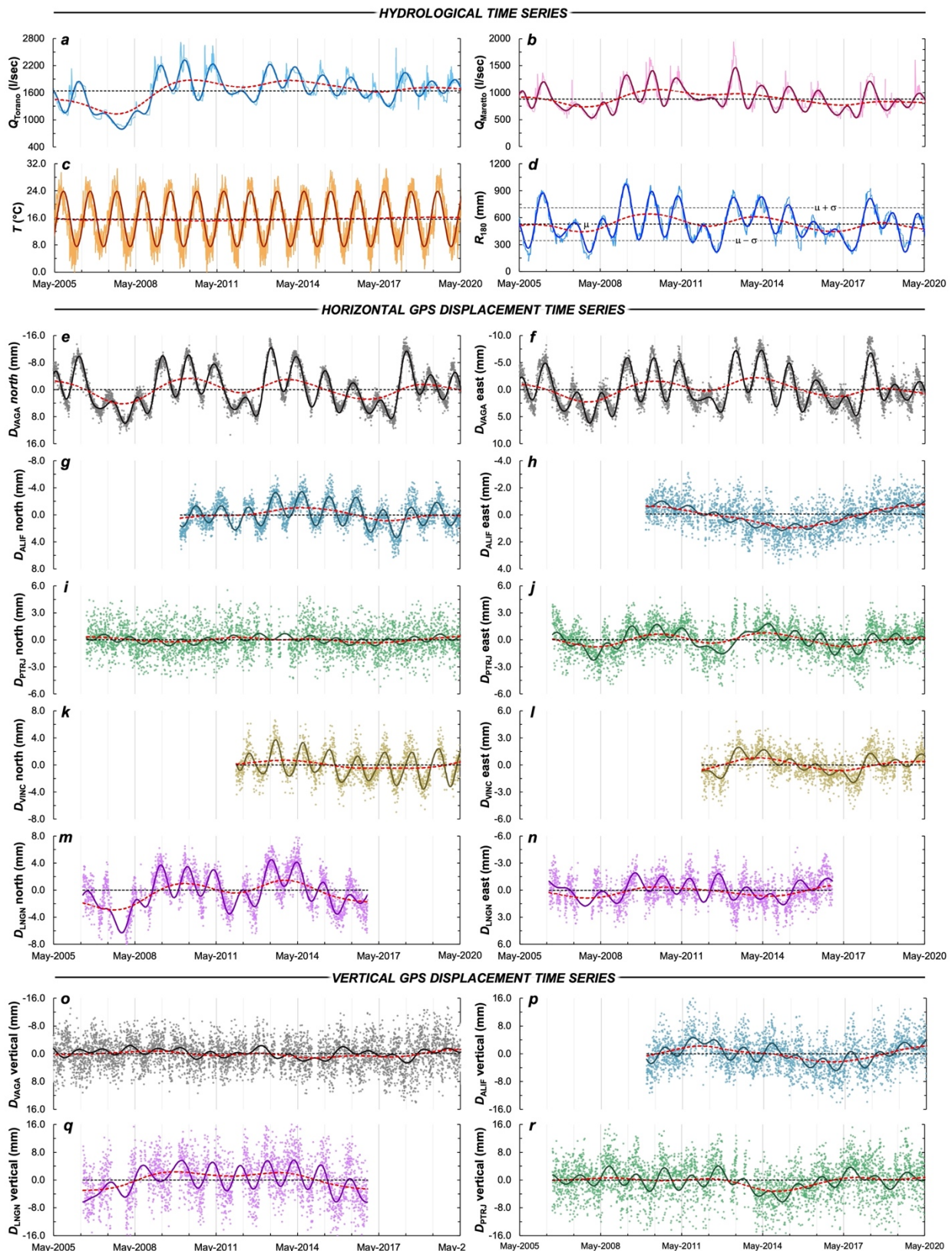


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 460 **Figure 1** Long-term time series of Maretto karst spring discharge (170 m a.s.l., Matese massif), black line, and horizontal
 461 displacement of *VAGA* GPS station (736 m a.s.l.), blue dots (daily data). Colors highlight the wettest (2011-2012) the
 462 driest (2016-2017) hydrological years of the time series.
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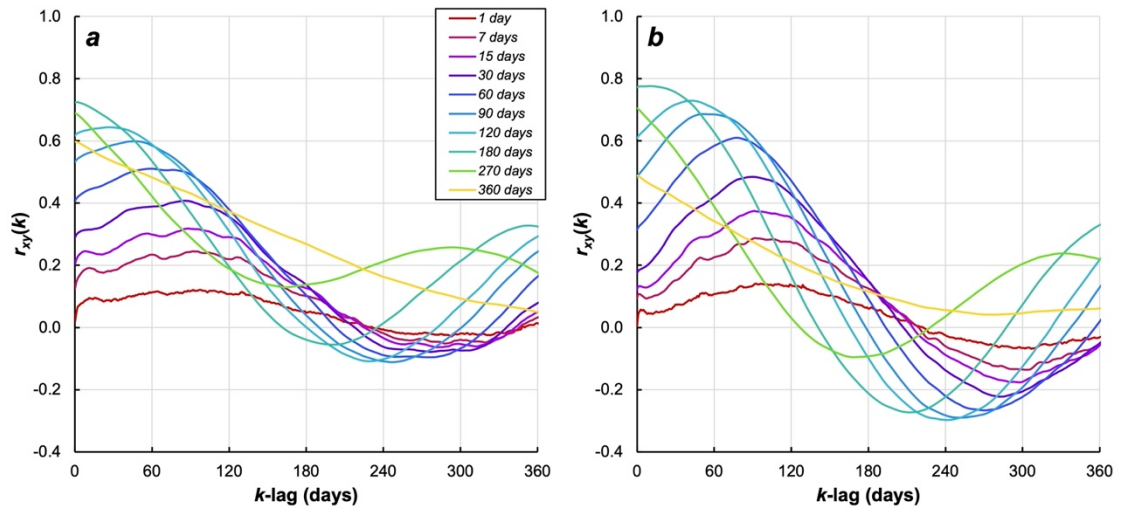
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Figure 2 Geological sketch of the Matese karst area. Legend: 1) slope breccias and debris, alluvial, lacustrine and travertine deposits (*Quaternary*); 2) pyroclastic deposits (*Quaternary*); 3) argillaceous complexes and flysch sequences (*Paleogene-Miocene*); 4) dolomite-limestone sequences (*Jurassic-Miocene*); 5) main karst spring (small circle symbol) and major karst basal spring (big circle symbol, with elevation in m a.s.l.); 6) boundary of the Matese karst massif; 7) internal runoff area; 8) GPS station (with elevation); 9) meteorological station (with elevation); 10) mountain peak (with elevation); 11) main village; 12) major river; 13) lake elevation; 14) trace of the hydrogeological cross-section shown in Figure 8.



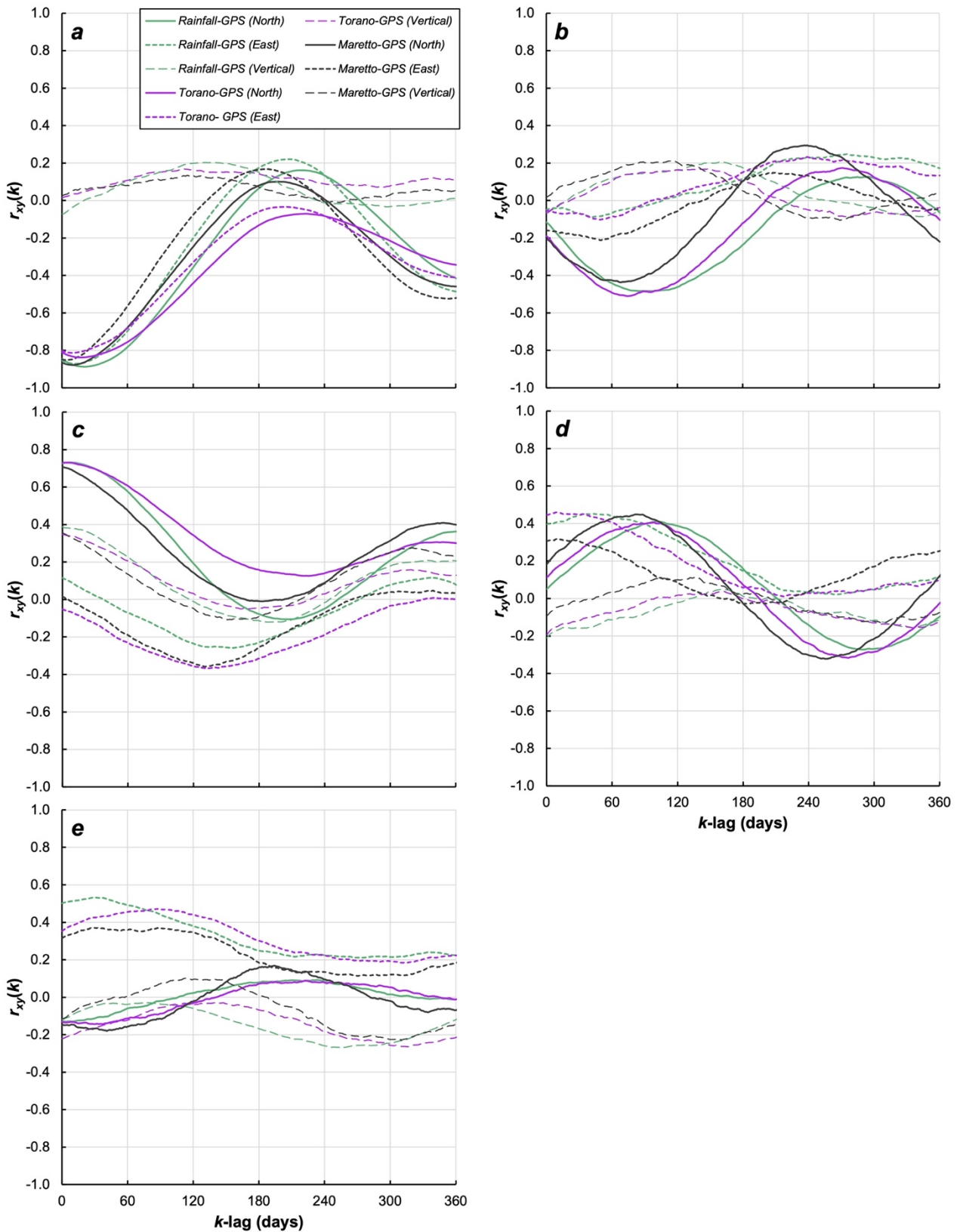
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474 **Figure 3** Hydrological time series of spring discharge (*a* and *b*), temperature (*c*), and 180-day cumulative rainfall (*d*),
 475 against horizontal displacement, D , north and east, time series for the different GPS stations (*e-n*). Note that the ordinate
 476 axis has been reversed in figures *e-f*, *g-h*, *n* and *o*. Black dashed lines indicate the time series mean (\square). Grey dashed lines
 477 in figure *d* indicate the mean plus and minus the standard deviation (σ). Smoothed colored curves were calculated by
 478 FFT-based analysis, while red dashed curves represent the 4-year Gaussian moving average.

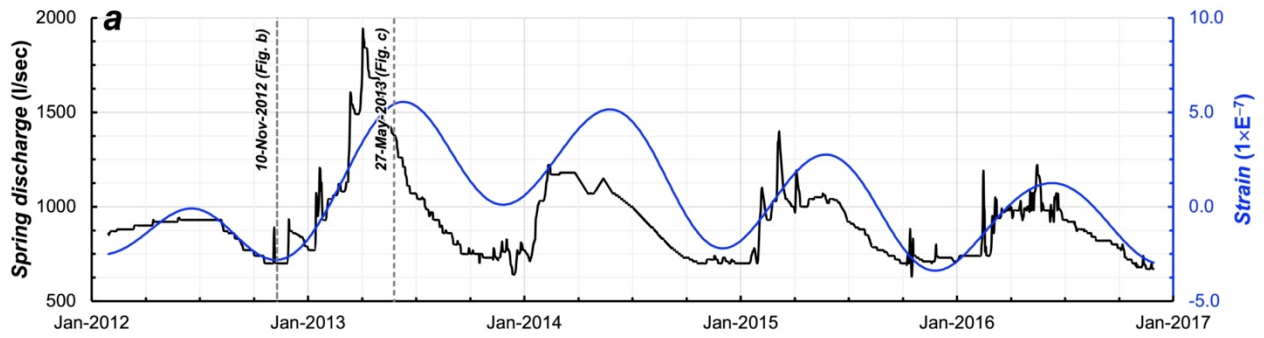


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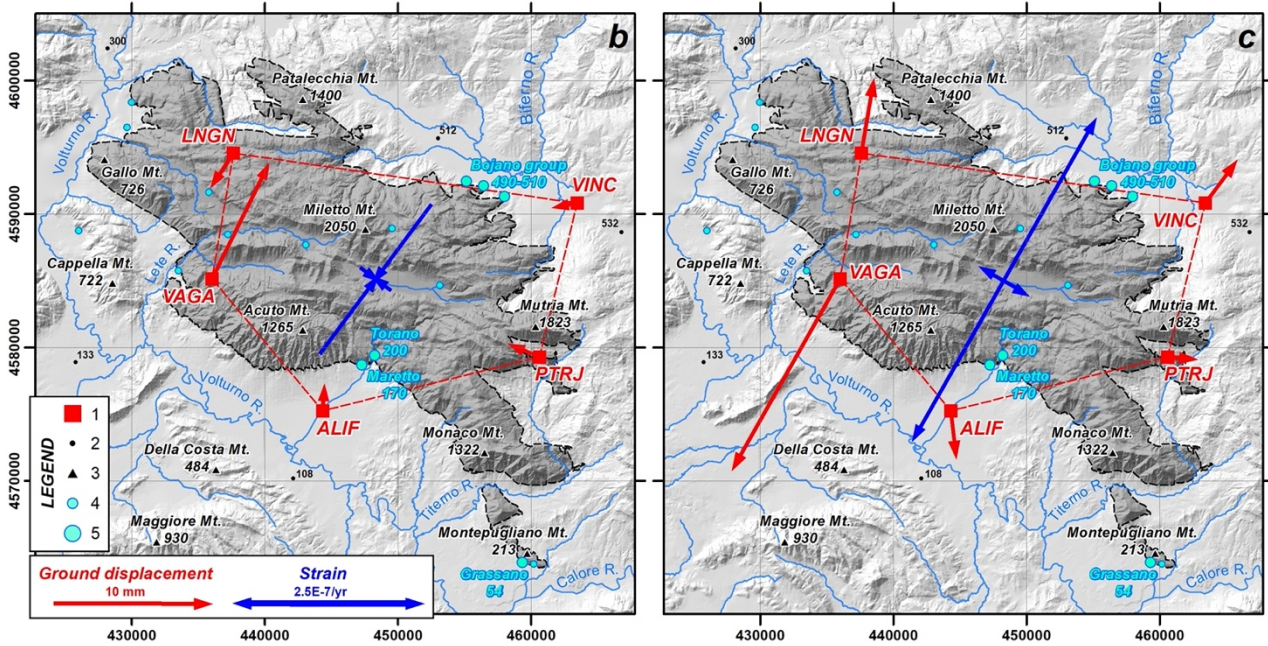
480 **Figure 4** Cross-correlation functions between rainfall cumulated over period of different length (Alife rain gauge, *S/I*
 481 station in **Figure 2**) and discharge of Torano (*a*) and Maretto (*b*) karst springs. Raw time series were used for cross-
 482 correlation analysis.



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 484 **Figure 5** Cross-correlation between hydrological records (rainfall, and Torano and Maretto spring discharge; time series
 485 of x) and displacement data of five GPS stations (lagged time series of y): *VAGA* (a), *ALIF* (b), *LNGN* (c), *VINC* (d),
 486 *PTRJ* (e). Note that time series of 6-month rainfall was used for panels from a to d, and 9-month rainfall for panel e. Raw
 487 time series were used for cross-correlation analysis.
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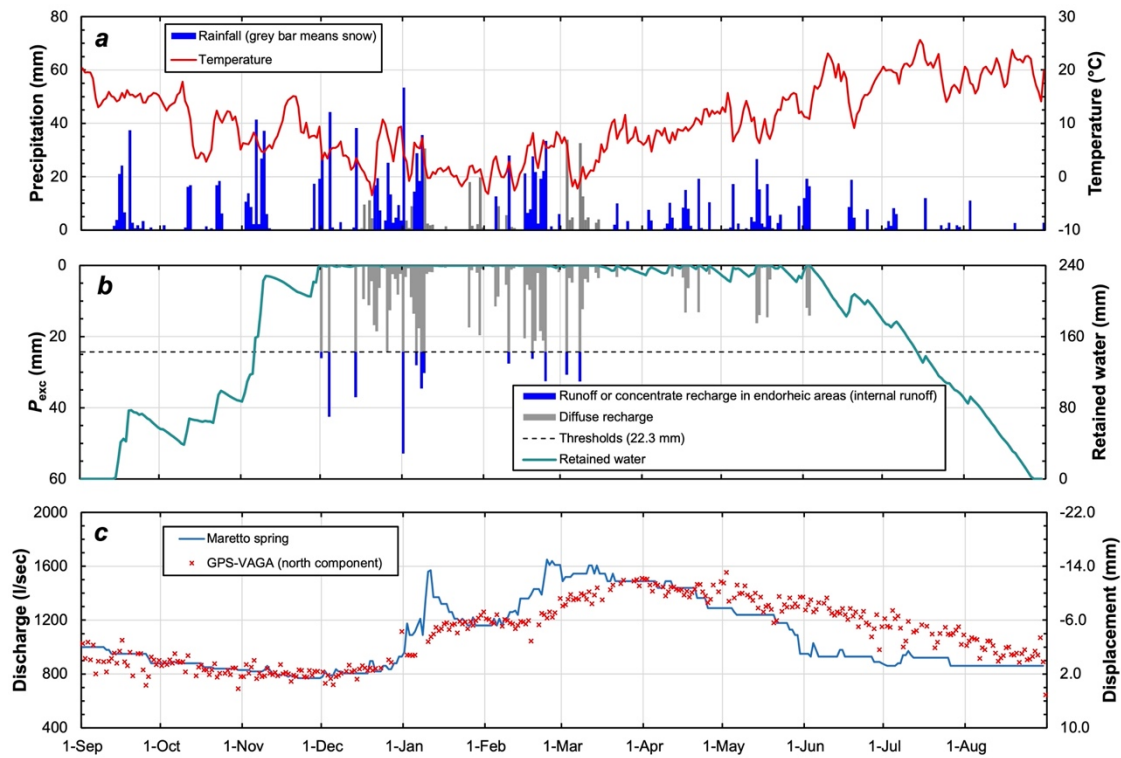


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491 **Figure 6 (a)** Time series of Maretto karst spring and temporal evolution of the horizontal estimated strain ($\epsilon_{xx} + \epsilon_{yy}$) in
 492 the polygon joining stations in the Matese massif (red dashed line in panels **b** and **c**), highlighting the strong correlation
 493 between horizontal deformation and hydraulic conditions in the Matese aquifer; positive strain values indicate dilatation,
 494 while negative values indicate contraction. GPS displacements (red arrows) and strain estimated (blue diverging and
 495 converging arrows) for 2 snapshots representative of high (**b**) and low spring discharge conditions (**c**; see the
 496 Supplementary Material for the continuous evolution). The Matese karst massif was emphasized with a grayish tone.



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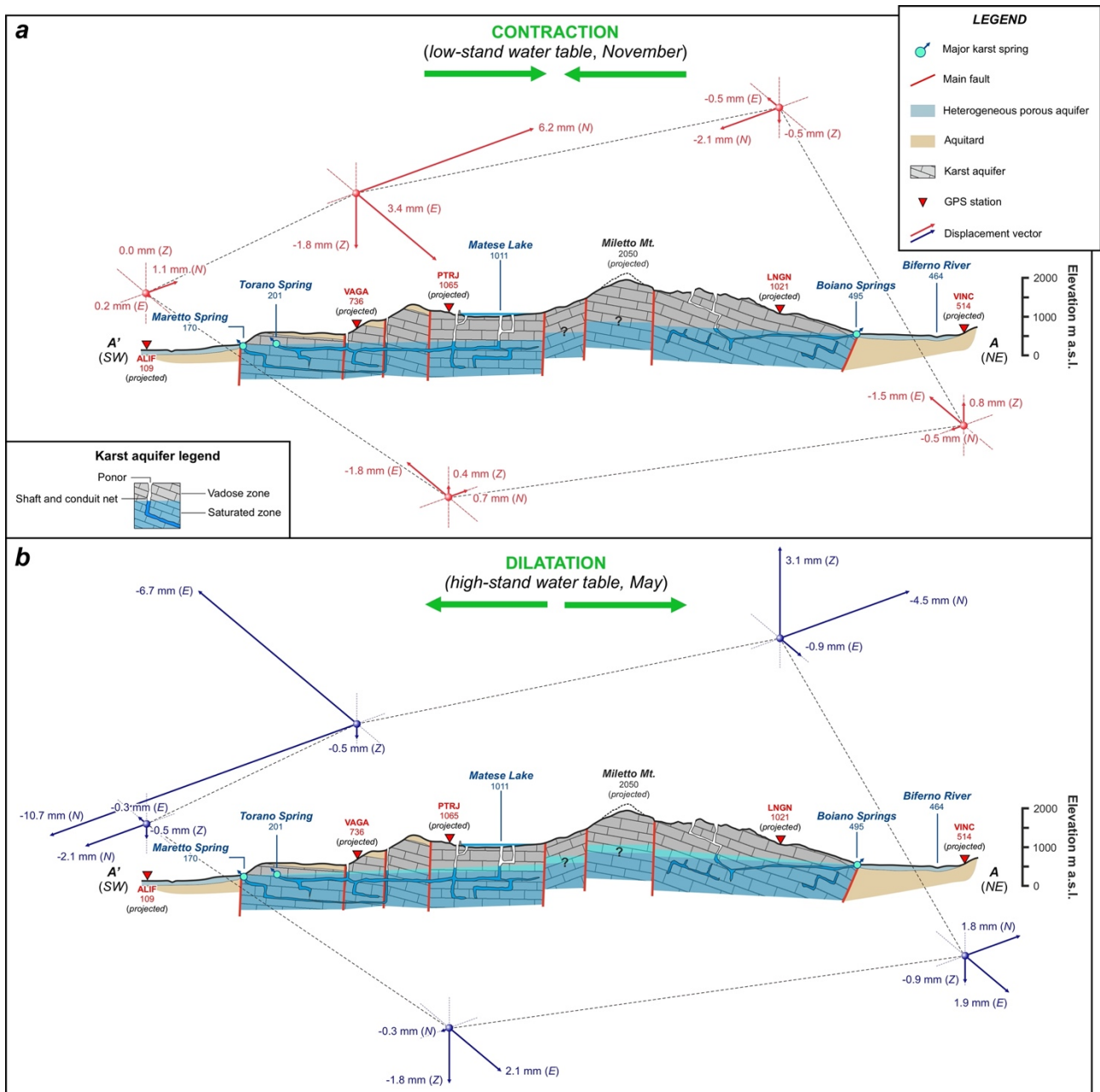
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Figure 7 Daily recharge model applied for the 2009-2010 hydrological year. Panel *a* shows rainfall, snow, and temperature records of Letino climate station (*S2*), located at 1050 m a.s.l. in western sector of the Matese massif. Panel *b* illustrates the simulated precipitation excess, P_{exc} , which was split into recharge and runoff (or concentrate recharge in endorheic areas) by a rainfall threshold value of 24.3 mm; simulated retained water was also shown, which was obtained from soil hydrological balance. Panel *c* shows discharge of Mareto spring and observed GPS displacement of *VAGA* station (north component).



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Figure 8 Hydrogeological sketch of the Matese karst massif (modified from Fiorillo and Pagnozzi 2015; section trace is shown in **Figure 2**) illustrating two different stage of the water table height (November 2012 and May 2013). Groundwater conditions in dry season (low flow; **a**) and wet season (high flow, **b**) are shown, along with components of the vertical (Z) and horizontal displacements (N and E) recorded by GPS stations.