

1 **Investigation of the climate-driven periodicity of shallow groundwater level fluctuations in**
2 **a Central-Eastern European agricultural region**

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18

19 **Abstract**

20 The distribution and amount of groundwater, a crucial source of Earth's drinking and irrigation
21 water, is changing due to climate-change effects. Therefore, it is important to understand
22 groundwater behavior in extreme scenarios, e.g. drought. Shallow groundwater (SGW) level
23 fluctuation under natural conditions displays periodic behavior, i.e. seasonal variation. Thus, the
24 study aims to investigate (i) the periodic behavior of the SGW level time series of an agriculturally
25 important and drought-sensitive region in Central-Eastern Europe – the Carpathian Basin, in the
26 north-eastern part of the Great Hungarian Plain, and (ii) its relationship to the European
27 atmospheric pressure action centers. Data from 216 SGW wells were studied using wavelet
28 spectrum analysis and wavelet coherence analyses for 1961-2010. Locally, a clear relationship
29 exists between the absence of annual periodic behavior in the SGW level and the periodicity of
30 droughts, as indicated by the self-calibrating Palmer Drought Severity Index and the Aridity Index.

31 During the non-periodic intervals, significant drops in groundwater levels (average 0.5 m) were
32 recorded in 89% of the wells. This result links the meteorological variables to the periodic behavior
33 of SGW, and consequently, drought. On a regional scale, Mediterranean cyclones from the Gulf of
34 Genoa (northwest Italy) were found to be a driving factor in the 8-year periodic behavior of the
35 SGW wells. The research documents an important link between SGW levels and local/regional
36 climate variables or indices, thereby facilitating the necessary adaptation strategies on national
37 and/or regional scales, as these must take into account the predictions of drought-related climatic
38 conditions.

39 **Keywords:** climate change, groundwater periodicity, Hungary, PDSI, wavelet spectrum- and
40 coherence analyses

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42 **NOTE TO COPYEDITOR – PLEASE INSERT THE FOLLOWING AS A FIRST-PAGE**
43 **FOOTNOTE:**

44 This article is part of the topical collection “Climate-change research by early-career
45 hydrogeologists”

46

47 1. Introduction

48 It is a widely accepted fact that water is one of the most critical (UNESCO 2012) and
49 threatened natural resources, and as such it is affected to a great extent by global and regional
50 climate change (Aeschbach-Hertig and Gleeson 2012; Vörösmarty et al. 2000). In this sense
51 groundwater is even more important because it represents, by volume, the second largest accessible
52 fresh water supply, after frozen water (Aeschbach-Hertig and Gleeson 2012). In addition,
53 groundwater is less vulnerable to droughts and anthropogenic contamination than surface water. It
54 provides about 40% of the world’s water used for irrigation (Siebert et al. 2010), and a fair amount
55 of the drinking water supply (Oki and Kanae 2006). Almost half of the world’s population depends

56 on groundwater as a source of drinking water, while due to water stress, more than 2 billion people
57 are exposed to a lack of potable freshwater and water for proper irrigation (Oki and Kanae 2006).

58 Although it is anticipated that in many regions climate change is set to increase the amount
59 of renewable freshwater resources still, the probability of extreme events is also expected to
60 increase (Oki and Kanae 2006), along with a general decrease in water levels. In certain regions,
61 the groundwater-level decline is related to substantial aquifer dewatering (Hoque et al 2007; Ta'any
62 et al 2009), while the increase in temperature due to climate change also has a significant driving
63 effect on the decrease in shallow groundwater levels (Chen et al 2004). To complicate the picture
64 further, in parallel with this overexploitation of water resources, water infiltration from
65 anthropogenic sources may locally reverse the general decrease and cause an increase in shallow
66 groundwater levels (Ta'any 2009). One indicator of the intermittent presence/intensity of these
67 driving forces is the absence of natural periodic behavior (i.e. seasonal variation) in groundwater
68 level fluctuations (Kovács et al. 2004). Taken together, these adverse scenarios directly affect both
69 the natural environment (Sophocleous 2000) and the consumers of water, humans included. The
70 effect on communities via drinking water is clear, but agriculture, and therefore those engaged in
71 agriculture, may also be affected, since these phenomena increase the cost of pumping and/or the
72 probability of wells drying up (Shah 2007). Adaptation to these unexpected changes is essential
73 (Green et al. 2011).

74 The analysis of groundwater level patterns has a long history in academic literature. To take
75 two recent examples, groundwater level patterns have been assessed on a global scale in terms of
76 groundwater availability (Fan et al. 2013), and used in order to see the possible impacts of climate
77 change as well (Green et al. 2011). Unfortunately, the available direct methods of measuring
78 groundwater levels are more suited to regional analysis than to analysis on a global scale.
79 Therefore, models developed for regional studies have to serve as stepping stones towards a
80 broader understanding of the behavior of groundwater level fluctuations. These models concern (i)

81 macro-regions with the assessment of phenomena in relation to global circulation models (Kuss
82 and Gurdak 2014; Holman et al. 2011) and/or more localized climatic variables (Chen et al. 2004),
83 and (ii) the derivation of models used in forecasting as well (Adamowski and Chan 2011; Mackay
84 et al. 2015; Chen et al. 2002). From a practical point of view, a common characteristic of these
85 macro-regional models is the use of state-of-the-art statistical tools, such as artificial neural
86 networks, wavelet spectrum/coherence analyses, etc. In combination with a broader perspective, in
87 the search for climatic driving factors of macro-regional or regional groundwater level fluctuations,
88 these tools are both necessary and may be expected to lead to an enhanced understanding of annual
89 and multi-annual scales of the groundwater level fluctuation prevailing in regional agricultural
90 areas and their vulnerability.

91 The aim of the present study was, therefore, to investigate (i) the periodic behavior of the
92 shallow groundwater level time series of an agriculturally important region in Central-Eastern
93 Europe which is highly vulnerable to arid conditions; and (ii) its supposed relationship to the
94 European atmospheric pressure action centers. If it is possible to draw a clear quantifiable
95 connection between the periodic characteristics of the pressure action centers and the recurrence of
96 drops in shallow groundwater levels in the study area, this work could contribute to management
97 strategies that are used to prepare for future droughts on a broader spatial scale.

98

99 2. Materials and Methods

100 2.1. Site description

101 The study area is located in the north-eastern part of the Great Hungarian Plain within the
102 Carpathian Basin in Central-Eastern Europe (47-48.2°N – 20.7-22.2°E; Fig. 1). The area is
103 characterized by intense agricultural activity (Burgerné Gimes 2014) which has an important role
104 for the whole Central-Eastern European region. Before the political and economic changes in the

105 former Soviet satellite countries (1989-1990), Hungarian agriculture was booming, and it was the
106 country's most successful economic sector. 60% of the investigated time span falls within that
107 flourishing period. Hungarian agriculture in 1989-1990 produced 17% of the total Gross Domestic
108 Product (GDP), providing work for about 22% of the total national labor force. Of the total amount
109 of export products, 22% came from the agricultural sector. These proportions, however, fell
110 significantly after 1990, reaching 3.3% (GDP), 4.7% (labor force), and 7% (exports), respectively
111 by 2007 (Burgerné Gimes 2014).

112 The climate of the study area is continental with an annual mean temperature of 10.4 °C and
113 555 mm yr⁻¹ mean precipitation (Fig. 2a) for the years 1961-2010. Regarding the annual
114 distribution of precipitation, the maximum occurs in early summer (i.e. June); a second maximum
115 can also be identified in late autumn, due to the impact of Mediterranean weather systems. The
116 precipitation minimum occurs in winter (mostly February); however, the potential risk of drought
117 generally appears in late summer because of higher summer temperatures (Fig. 2a). In the case of
118 any one year, drought might occur in any month. For instance, in the years where drought was
119 clearly recognized, such as years 1971-1974, arid conditions were present in spring and late
120 summer (Fig. 2b), and for those years there was 63 mm less annual precipitation compared to the
121 average for 1961-2010.

122
123 Although the area is not generally exposed to severe drought (Fig. 2), even a slight shortfall
124 in precipitation (one or two consecutive arid years) can result in extensive dry conditions and a
125 notable drop in shallow groundwater levels, as the soil is mainly sandy. In relation to this, about
126 60% of the area is characterized by downward seepage. Field experiments in the vicinity of the
127 study area have proven that in more consolidated soils, even 60 mm of rain can be taken up by the
128 soil without having a significant impact on the shallow groundwater levels within the span of two

129 days (Varga Cs. Expert, Nitrogen Works Co. Ltd., personal communication, 2016). According to
130 a recent study (Tóth 2012), one third of the area is exceptionally good for agricultural activity,
131 though from another perspective, another third is highly sensitive to environmental changes in
132 terms of their impact on potential agricultural production.

133 A decrease in SGW level was observed in the study area, leading to an increase in irrigation
134 water demand thus a further decrease in SGW levels; a similar finding was reported for the Danube-
135 Tisza interfluvium (southern Hungary) in a study by Szalai (2011). In the study area reported here,
136 the water for irrigation was first taken from the shallow strata (~15 m), then as the groundwater
137 levels dropped and the amount of available water decreased, deeper strata were tapped (~40 m).
138 The result was a depletion of the subsurface water reservoir. This resource shortage, however, was
139 hard to monitor because the amount of water used by the farmers can only be tracked by the
140 voluntary declarations of the farmers, and no inspections were yet conducted to record the situation.

141 It is important to note that, on the “sensitivity of environmental areas” scale, this area is
142 categorized as “highly sensitive” with respect to water protection and its natural environment when
143 compared with the country as a whole (Harsányi et al. 2013). In order to assess the phenomena
144 driving the arid conditions and climate in the study area, not only local weather conditions were
145 considered, but all of the key atmospheric pressure action centers affecting the weather of the
146 Carpathian Basin (as defined by Péczely, 1961) were taken into account: the Genoa Low (off the
147 coast of northwest Italy), the Icelandic Low, the Azores High, and the Eastern European High (for
148 details see **Section** ‘*Meteorological data*’).

149

150 2.2. Shallow groundwater monitoring wells

151 Due to the long history of agriculture in the area, the study had at its disposal the water-level
152 time series data of a dense and highly developed network of shallow groundwater monitoring wells,

153 giving a high degree of spatio-temporal coverage. Thus, 216 shallow groundwater level monitoring
154 wells, part of the Hungarian National sampling network, could be analyzed for the period 1960-
155 2010 (Fig. 1). The number of available wells increased continuously over time with a random
156 presence of missing data; the period with the greatest number of data gaps was the late 1970s.

157 The sampling frequency also varied over time as well as in space, e.g. in the 1960s the
158 sampling frequency was 3 days, while in 1970-1990 it was weekly, and post-1990 digital recording
159 increased the frequency to every 4 hours. By the beginning of the 2000s the dataset could be
160 categorized in three classes: 1-week, 3-day, and 4-hour data. The preprocessing steps of the data
161 are discussed in **Section** '*Statistical tools used*'.

162

163 2.3. Meteorological data

164 The meteorological data were acquired from the CarpatClim (Climate of the Carpathian
165 Region) online database (Spinoni et al. 2015) for the study area with a 0.1° horizontal resolution
166 and for the time interval 1961-2010. Specifically, mean air temperature, precipitation, cloud cover,
167 global radiation, relative humidity, and potential evaporation were included as the independent
168 meteorological parameters, as were the following indices: the standardized precipitation (SPI),
169 standardized precipitation-evapotranspiration (SPEI), self-calibrating Palmer Drought Severity
170 (scPDSI) and aridity (AI). However, of the mentioned indices, only the scPDSI and the AI proved
171 capable of showing an interpretable relationship with the periodic behavior of the shallow
172 groundwater levels in the area (for details see **Section** '*Wavelet Spectrum analysis*'). Therefore,
173 the computational background to just the scPDSI and AI will be presented in the following
174 paragraphs.

175 The scPDSI is derived from monthly precipitation, temperature and soil moisture data, taking
176 the local climate into consideration (Wells et al. 2004). Negative and positive values indicate dry

177 and wet conditions, respectively, with greater values implying drier/wetter climatic conditions. The
178 AI is the ratio of monthly precipitation and potential evapotranspiration; it basically indicates dry
179 and wet periods (Mihic et al. 2013).

180 Information concerning large-scale meteorological patterns and events, i.e. pressure data
181 driving the European large-scale weather phenomena affecting the Carpathian Basin (including the
182 study area), is represented from the gridded reanalysis pressure fields of the European Centre for
183 Medium-Range Weather Forecasts ERA-20C database (Hersbach et al. 2015; Poli et al. 2016),
184 which covers the entire 20th century and the first decade of the 21st century. The Gulf of Genoa, the
185 Icelandic Low and the Eastern European High are the key action centers which influence the
186 weather of the Carpathian Basin to the greatest degree. The average monthly sea-level pressure at
187 the representing grid points was downloaded from the online database of ECMWF.

188

189 2.4. Statistical tools used

190 As a first step, the data were preprocessed if the gap between any two data was smaller than
191 one month; the time series was interpolated by a cubic spline and resampled to 14 days to ensure
192 equal spacing without missing data, because this is one of the most basic requirements of
193 periodicity analyses methods. However, if the gap was longer than a month, then the time series
194 was split, and sections at least 10-yr long were obtained. It is well-known that any interpolation
195 will cause spectral bias and result in a “reddened” interpolated time series (Schulz and Stattegger
196 1997). The frequency band of the investigations (annual period) is not, however, one in which
197 spectral bias due to the current interpolation is to be expected. From the 57 time-series that are
198 longer than 30 yrs, using 20-yr low-pass filtering, the long-term trend was removed.

199 In the study, wavelet spectrum analysis (WSA) was used to assess the periodic behavior of
200 the shallow groundwater levels. The basis for WSA is the continuous wavelet transform ($W_n(s)$),

201 which could be defined as the convolution of the data X_n with a scaled and translated version of
202 the wavelet function ψ (Torrence and Compo, 1998; Eqn (1)), which is localized in both frequency
203 and time with a zero mean (Grinsted et al. 2004).

$$204 \quad W_n^X(s) = \sqrt{\frac{\Delta t}{s}} \sum_{n'=1}^N X_{n'} \psi^* \left[(n' - n) \frac{\Delta t}{s} \right] \quad (1)$$

205 Whereby the asterisk (*) represents the complex conjugate, s represents the wavelet scale, Δt
206 is the uniform time steps, and n is the localized time index. The Morlet mother wavelet (Morlet et
207 al. 1982) is the source function used to generate daughter wavelets by transforming them (Kovács
208 et al. 2010a). The Morlet mother wavelet was chosen because, with respect to other functions
209 (wavelet and e.g. sine), it describes the shape of hydrological signals quite well, while providing a
210 good balance between time and frequency localization (e.g. Gaucherel, 2002; Labat, 2005; Kang
211 and Lin, 2007). The aim of the wavelet transformation is multiple dissociation by decomposing the
212 data in the scaling space, thus making it possible to reveal its self-similarity structure (Hatvani
213 2014); for an example of the output of WSA see Fig. 3.

214 WSA provides the basis for wavelet transform coherence analysis (WTC). WTC is able to
215 indicate the areas with a common power of the two variables in the time-frequency space (Torrence
216 and Webster 1999). This was particularly important when the relationship of the periodic behavior
217 of the SGW wells and the climate data was assessed on a regional scale (see **Section 'Regional**
218 *driving factors'*)

219 A close resemblance can be observed between the WTC and the traditional correlation
220 coefficient. Thus, the WTC can be interpreted as the localized squared correlation coefficient in
221 the time-scale plane (Grinsted et al. 2004). However, the fact that WTC indicates a strong common
222 periodic behavior does not directly mean that the assessed two time series will correlate as well to
223 the same degree, because the periodic components must be present in both to show coherence in

224 the WTC output (Kern et al. 2016). Because wavelets are not completely localized in time, WTC
225 produces edge artifacts, thus the introduction of a cone of influence (COI) is suggested in which
226 edge effects cannot be ignored (Torrence and Compo 1998).

227 In the course of the evaluation, only those signals were taken into account which were
228 significant ($\alpha=0.95$) and the “phase information” could be interpreted (i.e. the signals were in-
229 phase, in antiphase, or the phase difference could be related to natural phenomena; for details see
230 Rösch and Schmidbauer 2014a). Moreover, spectral constraints were established by combining
231 WTC and band filtering, thus, the signal-to-noise ratio was successfully improved in certain cases
232 by extracting “focus” bands with a rectangular window.

233

234 2.5. Software used

235 R statistical environment (R_Core_Team 2008) was used to perform the calculations –
236 specifically, the WSA was done with the `analyze.wavelet` function – while the WTCs were
237 generated with the `analyze.coherency` function of the Wavelet-comp package (Rösch and
238 Schmidbauer 2014b) and the frequency filtering was done using the `astrochron` package (Meyers
239 2014). In addition, Surfer 11, QGIS 2.8, Excel 2016 and CorelDRAW X6 were used in the
240 preparation of the paper.

241

242 3. Results

243 The first step was to construct a dataset consisting of the time series of the 216 shallow
244 groundwater wells. Then, with WSA, their periodic behaviors were assessed. Consequently, the
245 time periods were extracted in which most of the wells indicated the absence of annual periodic
246 behavior. This pattern was compared with the local climate data and assessed statistically – using
247 wavelet coherence analysis – with the regional climate data (Fig. 4).

248

249 3.1. Annual periodicity

250 3.1.1. Wavelet Spectrum analysis

251 As indicated by WSA, clear peaks can be observed in a number of wells with missing annual
252 periodicity (Fig. 5a). The absence of the annual periodicity was not observed in all the wells at the
253 same time. There were four distinct time intervals in which the one-year period was missing in
254 more than 50% of the wells and four additional ones if this boundary is lowered to 25% of the total
255 number of available wells. The most characteristic interval when the annual periodicity was absent
256 peaked in 1974, when more than 90% of the wells did not show annual periodicity. From this point
257 on, the gaps in annual periodicity decreased regarding both the number of wells involved and the
258 duration of the time periods. If the number of wells in which annual periodicity was absent reached
259 25% of the total number available at that particular time, this was considered a *1-yr absence peak*,
260 where the term “1-yr” refers to the absence of the annual periodicity. The five 1-yr absence peaks
261 sub-divided the full time interval into 11 sections, where periodic and non-periodic time intervals
262 follow each other in tandem (Fig. 5a). If the peaks are assessed using WSA, a significant 8-yr return
263 cycle can be determined (Fig. 5b).

264 WSA was systematically conducted on all the meteorological parameters that were available
265 and could possibly have been driving the periodic behavior of the shallow groundwater levels.
266 However, either there was no absence of annual periodicity in the meteorological parameters’ time
267 series in the investigated time at all (e.g. mean air temperature, global radiation, cloud cover,
268 relative humidity, potential evaporation), or the period with no annual periodicity (e.g.
269 precipitation, SPI, SPEI) did not overlap the periods with no periodicity in the shallow groundwater
270 wells.

271 Since groundwater level depends both on the available water income (i.e. precipitation) and
272 water loss (i.e. evaporation), which is also highly affected by temperature, complex climatic indices
273 are more efficient in reflecting the statistical behavior of SGW level time series. Therefore, the AI
274 (Fig. 6a) and the scPDSI (Fig. 6b) were selected from the available meteorological variables.

275 The results of the AI with respect to annual periodicity show a similar behavior to the
276 groundwater level in the SGW wells, namely, the 1-yr absence peaks of AI were mostly in line
277 with the gaps of annual periodicity in the SGW wells (Fig. 6a).

278 Because in the derivation of the scPDSI, a 7-8-month average of potential evapotranspiration
279 values is used (Wells et al. 2004), WSA was not able to locate the annual periodicity. However, as
280 expected, the visual comparison between the 1-yr absence peaks and the negative peaks in the
281 scPDSI time series concurred in almost all cases (Fig. 6b). The most characteristic concurrence can
282 clearly be identified in 1974, when the scPDSI is below -4, indicating an extreme drought. The
283 negative scPDI peaks before 1966 and after 2008 (Fig. 6b) were not reflected in the 1-yr absence
284 peaks because of the edge effects causing high uncertainty (Torrence and Compo, 1998) in the
285 wavelet spectra of the SGWs.

286

287

288 4. Discussion

289 4.1. Local driving factors

290 The absence of annual periodicity in the SGW wells is also known to have occurred in other
291 areas of Hungary since the mid-2000s (Kovács et al. 2004, Szalai 2011). However, detailed
292 investigation of the study area only began around 2010 (Kovács et al. 2010b). The absence of
293 periodic behavior determined in these studies was significant with regard to the annual, 5-yr, and
294 11-yr period bands. It should be noted, however, that the methodology used by Kovács et al. 2004

295 was (i) only capable of determining whether the absence occurred or not, and (ii) was not able to
296 localize the phenomena in time, unlike in the present study using WSA.

297 The presence or absence of periodic behavior in the SGW levels is in and of itself not
298 informative enough to prepare for future droughts and/or the impact of climate change without
299 knowledge concerning local and regional driving factors. A relationship between standardized
300 precipitation indices and groundwater drought return periods (Kwak et al. 2013), as well as
301 groundwater drought in general, has been uncovered previously (Bloomfield and Marchant 2013;
302 Khan et al. 2008; Kumar et al. 2016; Raziei et al. 2009). Although numerous indices have been
303 examined in the present study, whether or not these are linked to SGW periodicity, their
304 relationships were not present or at best inconclusive. As seen in **Section 'Wavelet Spectrum**
305 *analysis*' the most comprehensive link between local meteorological conditions and the periodic
306 behavior of SGW was found using scPDSI and AI. As far as can be ascertained from the literature,
307 the relationship between the periodic behavior of (i) SGW level and (ii) scPDSI and AI as measures
308 of drought, has not been thoroughly investigated before, especially not with a special focus on the
309 absence of periodic behavior. Only a couple of studies exist on this topic, e.g. Edossa et al. (2015),
310 who investigated the question of whether a relationship exists between the periodic behavior of
311 SGW level fluctuations and scPDSI in Africa, or Chen et al. (2002), who determined 7-8 and 13-
312 14 yr periods in the hydrographs, but did not link them to local or regional phenomena.

313 In the study area here, the connection between drought indicated by scPDSI and AI, and the
314 periodic behavior of SGW levels manifested itself directly in the water levels of the wells. Based
315 on the results, if periodic behavior of the water levels in the wells changes and/or is lacking, it is
316 to be expected that recharge from precipitation will have decrease as well, since it was found that
317 during the 1-yr absence peak periods (Fig. 5a), the average water levels of the wells were indeed
318 lower than that which was to be found between them. This was true in 9 out of the 11 periodic and
319 non-periodic time intervals. During the five 1-yr absence peaks, the average shallow groundwater

320 level was 100.63 m, and 101.18 m between them, indicating a 55 cm difference for all the wells
321 available during the period spanning 1961-2010. This phenomenon can be investigated more
322 closely for each well separately and visualized for the study area using kriging (Cressie 1990,
323 Oliver and Webster 2014) (Fig. 7). For 89% of the wells available, it was true that on average the
324 water levels were lower during the 1-yr absence peaks than the average water levels between them.
325 The average water-level depth in the wells from the topography, i.e. ground surface (Fig. 7a; 1961-
326 2010), was compared to the average difference of water levels between the 1-yr absence peaks
327 (Fig. 5a) and the periodic years for each well separately (Fig. 7b). It became clear that the difference
328 between the water levels of the periodic and the non-periodic intervals was greatest (i.e. the drop
329 in water level was the greatest) in those areas where the water levels measured from the topography
330 were already generally deeper (the north-northeastern part of the study area). These areas overlap
331 with the agriculturally more exploited regions in the study area (Harsányi 2013), which seem to be
332 most vulnerable in this way. The drops in water levels between the periodic time intervals and the
333 1-yr absence peaks decrease the height of capillary saturation in the non-periodic time intervals,
334 calling for adaptation in terms of species- and crop-yield expectations. Meanwhile, the difference
335 was the smallest in the western part of the study area, where the water levels were, in general,
336 higher (Fig. 7a).

337
338 The areas with high SGW levels were the discharge areas (Székely 2003), while the ones
339 where drought – based on this study – has the greatest effect are the recharge areas (Székely 2003).
340 Thus, the link between the non-periodic time intervals and drought (low SGW levels) may be
341 related to the vulnerability of the recharge areas in the northeast (Fig. 7b).

342 According to Wilhite (1997) “drought is a normal, recurrent feature of climate that may occur
343 everywhere even though its characteristics and impacts vary significantly from region to region”.

344 To reveal the reasons or driving factors behind the substantial changes in local climatological
345 conditions, the investigation of the possible teleconnections between SGW periodicity and the
346 pressure action centers influencing the area would appear to be necessary.

347

348 **4.2. Regional driving factors – pressure action center in the Gulf of Genoa**

349 A complex analysis should include the large-scale factors which influence the climate of the
350 Carpathian Basin. The large scale weather patterns are dominated by pressure action centers on a
351 European continental scale (e.g. Barnston and Livezey, 1987). The main high pressure action
352 centers are located over Eastern Europe and near the Azores. Anticyclones often form over these
353 regions and result in calm weather conditions in their vicinity; however, they do not move as fast
354 as the midlatitude cyclones with low central pressure which result in substantial changes in local
355 weather conditions due to their frontal zones (e.g. Barry and Carleton, 2001). Low pressure action
356 centers are located near Iceland (aka the Icelandic Low) and in the Gulf of Genoa (Mediterranean
357 cyclones form here and approach the Hungarian study area). Although Mediterranean mid-latitude
358 cyclones are generally weaker (i.e. the central low pressure of the cyclone is generally less low,
359 and the cyclone size is smaller), with shorter lifetimes than the cyclones originating from the
360 Icelandic Low (Bartholy et al. 2006), their local effects on the weather phenomena and climatic
361 events of the Carpathian Basin are quite strong (e.g. Alpert et al. 1990), resulting in intense
362 precipitation and wet conditions in the target area of this study. This stronger relationship is
363 partially explained by (i) the shorter distance from the Gulf of Genoa to the study area compared
364 to the longer distance between the Icelandic Low and the study area, and (ii) the general patterns
365 of cyclone tracks within the southern part of Europe, which includes the Carpathian Basin too,
366 while the tracks of the Icelandic cyclone centers are usually north of the study area (Kelemen et al.
367 2015; van Bebbber 1891).

368 Thus, the relationship between the pressure center of the Gulf of Genoa and the study area
369 was proven to be most persistent and strongest out of the three pressure centers studied. The 8-yr
370 return period of the 1-yr absence peaks was compared with the spatial average pressure of the Gulf
371 of Genoa with WTC (Fig. 8a). To enhance the signal-to-noise ratio, the periods under 7 years were
372 removed from the spectrum of both time series, thus, the 8-yr common period was revealed to be
373 present to a significant degree, with the highest power between 1975 and 1995 (Fig. 8a), and in-
374 phase (Fig. 8b) throughout the whole investigated period. This means that when 1-yr absence peaks
375 occurred, the regional air pressure in the Gulf of Genoa reached its maxima, thus, the corresponding
376 occurrences of cyclogenesis in the region were substantially less frequent than usual. This
377 relationship can be explained by the fact that the occurrence of fewer cyclones results in less
378 precipitation and drier conditions along the cyclone paths, including this study area. Climatic
379 conditions are reflected in SGW wells, which can hence be considered as a good indicator of local
380 climatic change. Those large-scale climatic processes projected well by global climate models (for
381 which quite high confidence has already been built) enable the prediction of finer-scale
382 regional/local phenomena. These physically-based predictions should be considered as key input
383 information to develop appropriate adaptation and mitigation strategies on regional/local scales.
384 Due to the agricultural role of the study area, such strategies are especially important in reducing
385 the vulnerability of local economies.

386 In the past century-long period, only slight changes can be detected as a synoptic-scale
387 atmospheric impact of global climate change (Bartholy et al. 2009). Nevertheless, this century may
388 result in greater changes of cyclone activity and cyclone tracks. Projected future cyclone
389 climatology over Europe is analyzed by Muskulus and Jacob (2005) using a single climate model
390 and one specific scenario. Their results suggest more Mediterranean cyclones overall, but fewer
391 strong cyclones. The resulting precipitation is highly dependent on the strength of the cyclone; the
392 stronger the cyclone (i.e. the lower the central air pressure), the greater the amount of precipitation

393 along the cyclone track. However, a comprehensive assessment of future conditions would require
394 several climate models and different scenarios.

395

396 **5. Conclusions**

397 With the complex hydrogeological/meteorological model developed in this study, non-
398 periodic time intervals were found in the shallow groundwater levels (in 216 wells) of an
399 agriculturally important study area in Central-Eastern Europe. These statistical behaviors were then
400 proven to have a clear pattern of coincidence with drought indices (scPDSI and AI). The time
401 intervals with the annual periods missing in at least 25% of the available shallow groundwater
402 wells were (i) therefore linked to drought conditions, (ii) proven to have significantly lower water
403 levels mostly in the recharge areas of the studied sector, and (iii) linked to the maxima of the
404 regional air pressures in the Gulf of Genoa. In this sense, the study found an evident but as-yet
405 undocumented link between the (non)-periodic behavior of SGW levels and drought indices, and
406 also for the first time described a relationship between SGW periodicity and regional circulation
407 patterns. Thus, on the basis of the available climate model projections, estimated changes in the
408 large-scale circulation patterns provide key useful information concerning local climatic changes,
409 including possible changes in drought frequency, duration, and severity. Since drought-related
410 characteristics and their future changes strongly affect local agricultural production, necessary
411 adaptation strategies on either national or regional levels should take into account the predictions
412 of drought-related climatic conditions. If plant or crop species to be cultivated in a specific region
413 are selected according to the modified climatic conditions, many potential local problems induced
414 by global warming can be avoided.

415

416 **Acknowledgements**

417 The authors would like to thank the data providers: CARPATCLIM Database © European
418 Commission – JRC, 2013; ECMWF and the Water Directorates of the region. In addition, we are
419 grateful for the support of the MTA “Lendület” program (LP2012-27/2012), the János Bolyai
420 Research Scholarship of the Hungarian Academy of Sciences and the “Agrárklíma2” Project
421 (VKSZ_12-1-2013-0034). This is contribution No. 52 of 2ka Palaeoclimate Research Group. The
422 authors appreciate the contributions of the Early Career Hydrogeologists’ Network (ECHN) of the
423 International Association of Hydrogeologists (IAH) and the journal editorial team and reviewers.

424

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597

598 FIGURE CAPTIONS:

599

600 **Fig. 1.** The location of the study area in Central-Eastern Europe – the Carpathian Basin, in the
601 north-eastern part of the Great Hungarian Plain (the relief map is based on Horváth et al. 2006).
602 The parameters used in the study are the gridpoints of the ERA20C database representing the
603 pressure center in the Gulf of Genoa (*red crosses*, upper-left map), the shallow groundwater
604 observation wells (*red dots*, lower-right grid-map) and the gridpoints of the CarpatClim database,
605 representing the meteorological data (*grey crosses*, lower-right grid-map). [Codes for countries
606 adjacent to the study areas -- FR: France, CH: Switzerland, AT: Austria, SI: Slovenia, HR: Croatia,
607 RS: Serbia, RO: Romania, UA: Ukraine, SK:Slovakia]

608

609 **Fig. 2** Walter-Lieth diagrams for **a** 1961-2010 (annual mean temperature 10.4 °C and precipitation
610 555 mm yr⁻¹) and **b** 1971-1974 (annual mean temperature of 10.4 °C and precipitation 492 mm yr⁻¹). Both cases represent the whole study area

612

613 **Fig. 3** An example of the WSA output, with: **a** the water levels of a SGW well (*blue dots*); **b** the
614 20-yr low-passed 14-day resampled time series of the same well; and **c** its power spectrum density
615 graph. The 5% significance level against red noise is shown as a *thick black contour*. The *black*
616 *shaded areas* mark the cone of influence (COI). The *black dashed horizontal line* indicates where
617 the annual periodicity could be found (e.g. Jan 1961 – May 1970; Nov 1976 – Oct 1981)

618

619 **Fig. 4** Flowchart describing the steps of the analyses. WTC refers to wavelet transform coherence
620 and WSA to wavelet spectrum analysis

621
622 **Fig. 5 a** Summary figure for the number of wells where the annual periodicity was missing,
623 indicating the 1-yr absence peaks, where 100% equals the total number of wells available in that
624 particular year. **b** WSA result of the 1-yr absence peaks, where the *dashed line* indicates their 8-yr
625 return period

626
627 **Fig. 6 a** Summarized WSA result for the AI, in which the *horizontal lines* represent a given grid
628 point's time series, and **b** scPDSI time series compared to the 1-yr absence peaks. The *red ellipses*
629 indicate the concurrence of the 1-yr absence peaks and the minima of the scPDSI. If a given line
630 in **a** is red, the annual periodicity was present; if it is grey the annual periodicity was not present in
631 the grid point's time series.

632
633 **Fig. 7** Isoline maps of **a** average water depths from ground surface, and **b** average difference in
634 water levels between the 1-yr absence peaks and the periodic years for each well separately for
635 1960-2010, measured in meters. The maps were derived using ordinary point kriging with the
636 isotropic variogram models **a**: $C_0=0.001$; $C_0+C=1.986$; $a=14400$ m and **b**: $C_0=0.0052$;
637 $C_0+C=0.0494$; $a=5600$ m (for details see Oliver and Webster 2014)

638
639 **Fig. 8.** The WTC graphs of the 1-yr absence peaks and the monthly average air pressure in **a** the
640 Gulf of Genoa and **b** phase difference. *Horizontal black dashed line* indicates the 8-yr periodicity.

641