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Identifying the impact of climate and anthropic pressures on karst aquifers using wavelet analysis

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Abstract

This paper assesses the implications of climate and anthropic pressures on short to long-term changes in water resources in a Mediterranean karst using wavelet analysis. This approach was tested on 38-year (1974–2011) hydrogeological time series recorded at the Lez spring (South France), which is exploited for water supply. Firstly, we investigated inter-relationships in the frequency domain by cross-correlation across multiresolution levels. Our results showed that rainfall and spring discharge are highly correlated in the high frequency

domain which reflects the hydrogeological response during flood events of typical highly karstified systems. Pumping and groundwater level are correlated in a lower frequency domain, illustrating seasonal to multi-year relationships. Secondly, continuous wavelet transform was applied to characterize the temporal variability of the inter-relationships involved. On the contrary to examples of “non-managed” karst aquifers in the literature, our results showed that the 10-year rainfall component was attenuate in the discharge signal. We assume that the reason is that the storage variations are strongly affected by pumping. This interesting result shows that possible long-term impacts of rainfall variability due to climate change may be masked by a high pumping rate. We showed also that despite an increase of the pumping rate from the 1980s, the stress on the groundwater resource does not increase from year to year. The present pumping strategy does not affect the drawdown in the long term, avoiding an over-exploitation of the aquifer. Finally, this study highlights the effectiveness of wavelet analysis in characterizing the response variability of karst systems where the hydrogeological regime is modified by pumping.

Keywords

Karst; Wavelet transform; Multiresolution analysis; Hydrogeological processes; Climate and anthropic pressures; Signal processing

1 Introduction

Groundwater is a major global water supply resource and is currently affected by two main stressors: climate and anthropic pressures. This is essentially true for aquifers pumped for water supplies in Mediterranean areas due to increased abstraction to meet the needs of the growing population in regions where the aquifer is irregularly recharged from one year to the next. Evaluating the impacts of climate and anthropic pressure on water resources in such regions is a major challenge, as most large aquifers are located in carbonate rocks subject to karstification. The hydrogeological response of karst systems is highly non-linear due to spatial and dynamic heterogeneities linked to fact that the void structure leads to the formation of preferential drainage axes (for reviews see Bakalowicz, 2005; Goldscheider et al., 2007). Some of karst aquifers are an important water source for major cities, particularly in Mediterranean regions. In these cases, the aquifer may be referred to as “actively managed” if the pumping rate is higher than the low water stage discharge rate of the system under natural conditions in summer. Then, groundwater storage is highly mobilized before the rainy autumn period that contributes most of the annual recharge each year. In this paper, we investigate climatic and anthropic impacts on the groundwater resource in a Mediterranean karst system under active water management. We ask whether pumping modified the hydrogeological response.

Wavelet analysis has become a powerful technique to study geophysical processes or signals (Kumar and Foufoula-Georgiou, 1997; Torrence and Compo, 1998). Decomposing a time series into time-scale space, this method localizes power variations within a time series. It is ideal for analysing non-stationary signals and identifying short- to large-scale periodic phenomena. In the field of hydrology, Continuous Wavelet Transforms (CWT) have recently

64 been used to study the effect of climatic phenomena on the stream flow regime (Labat et al.,
65 2005; Massei et al., 2007; Labat, 2010; Fu et al., 2012), or to study runoff processes
66 (Lafrenière and Sharp, 2003; Schaefli et al., 2007). CWT has been widely used to study the
67 hydrogeological behaviour of karst systems. Comparing three springs, Labat et al. (2000;
68 2002) demonstrated the potential of wavelet analysis in identifying karst properties in relation
69 to the degree of karstification. Structural heterogeneity also determines similar filtering
70 properties on a small basin scale (Chinarro et al., 2011) and on a large scale (Hao et al.,
71 2012): i) short time-scale signals are generally less filtered showing the transmissive role of
72 the conduit network, ii) and high-energy large timescale signals can penetrate through the
73 aquifer, illustrating the buffered role of the storage zone. In a Mediterranean context, this
74 allows us to visualize annual and multi-annual scale components in relation to North Atlantic
75 Oscillation (Andreo et al., 2006). The hydrogeological response has also been studied from a
76 physico-chemical time series: to investigate transport properties and turbidity dynamics
77 (Massei et al., 2006), to highlight temperature-runoff relationships during snowmelt
78 (Mathevet et al., 2004), or to study groundwater variations in relation to the geological
79 context (Slimani et al., 2009). Surface-groundwater interactions were also studied using CWT
80 to improve understanding of river flow components in karst environments (Salerno and
81 Tartari, 2009). But, there is a lack of knowledge regarding the identification of the respective
82 role of climatic and anthropic pressures on the resource of karst aquifers.

83 Generally, CWT provides a good representation of energy distribution in time-scale
84 space in all these works, highlighting the non-stationary nature and multi-scale behaviour of
85 karst systems. However, to overcome limitations arising from intrinsic redundancy of the
86 CWT representation, Labat et al. (2000; 2001) applied an orthonormal wavelet representation,

87 conserving the signal information, as a complementary approach. This multiresolution
88 analysis (discrete wavelet transform DWT) can be used to decompose a signal into successive
89 resolution levels. It allows the energy distribution across levels to be characterized and the
90 slow and fast components in a spring discharge time series to be distinguished. This
91 complementary DWT technique allows easier and more efficient interpretation of the energy
92 distribution across decomposition scales, assisting the study of the time-frequency space of
93 time series. It can thus greatly improve hydrogeological understanding of karst systems, but is
94 rarely found in the literature. From these latest studies showing the significant potential of
95 combined CWT and DWT, we expect that both approaches will be adapted to characterize on
96 groundwater resources the response to the cumulative effect of climatic and anthropic
97 pressures.

98 To assess the impacts of climatic and anthropic pressure on groundwater resources in a
99 karst system under active water management, this study aims to distinguish between the role
100 of rainfall and pumping on the karst response using wavelet analysis. As a first step, a
101 multiresolution analysis was performed in order to characterize the energy distribution across
102 scales. To identify the frequency domain where rainfall and pumping may influence the karst
103 system, we also present a cross-correlation across multi-resolution levels. As a second step,
104 continuous wavelet analysis was applied to track changes in phenomena over time, providing
105 information on the temporal variability of the karst response to climatic and anthropic
106 stressors. We applied these techniques on rainfall, pumping, discharge and piezometric time
107 series over a 38-year period (1974–2011) in the Lez aquifer in the South of France.

108 **2 Site and measurements**

109 **2.1 Study site**

110 **2.1.1 Presentation**

111 The Lez karst is located to the north of Montpellier in the Cévennes area in the South of
112 France, in the western section of the Mediterranean zone. The Lez karst system is part of the
113 North MonPELLIÉRAINS karst hydrogeological unit bounded to the west by the Hérault River
114 and to the north and east by the Vidourle River (Figure 1). The Lez karst aquifer is located in
115 Upper Jurassic formations between 650 m and 800 m thick, located on both sides of the
116 Matelles fault. The aquifer is unconfined to the west of the fault, while the section located to
117 the east may be partially captive. In the zone lying under a Tertiary overburden, the aquifer is
118 found in the Upper Jurassic and the Lower Cretaceous. A more detailed description of the
119 study area can be found in Ladouche et al. (2014).

120 The Lez spring is the main outlet of the karst system (Figure 1). The spring outlet has
121 been explored by cave-divers. They discovered a huge saturated sub-horizontal karst conduit
122 developing more than 400 m inland (Figure 2), with a diameter ranging between 5 and 10 m.
123 The exploration ended at 113 m deep below the spring outlet (-48 m ASL) in a zone where the
124 conduit become wider. The hydrogeological basin is estimated to cover an area of 380 km²
125 (Thiery et al., 1983). Different recharge zones can be distinguished, depending on the nature
126 of the geological overburden. Recharge of the aquifer takes place predominantly in Jurassic
127 limestone, occupying an area of 80 to 100 km² (Figure 1). Within the Cretaceous overburden
128 (120 km²), losses occur locally along temporary watercourses and feed the aquifer locally
129 during flood events. The Tertiary formations occupy an area of about 160 km². In general

130 these are considered as impermeable or almost impermeable and do not contribute to
131 recharging the Lez karst aquifer.

132 The drinking water supply of the Montpellier agglomeration (with about 340 000
133 inhabitants) comes from the Lez karst spring since the 19th century (1854). Before 1968, this
134 resource was used by gravity extraction, varying between $25 \cdot 10^{-3}$ and $0.6 \text{ m}^3/\text{s}$ (Paloc, 1979).
135 From 1968 to today, the Lez karst spring is pumped according an active management strategy,
136 the pumping flow rate during summer periods is greater than the spring's low-water discharge
137 so as to mobilize the aquifer's stored reserves (Avias, 1995). From 1968 to 1982, water was
138 abstracted by pumping in the Lez Spring basin (Figure 2a) at a rate of the order of $0.8 \text{ m}^3/\text{s}$.
139 From 1983 onwards, deep boreholes located in the main karst conduit located upstream from
140 the spring (Figure 2) have allowed pumping at a rate of up to $1.7 \text{ m}^3/\text{s}$ (Avias, 1995). The
141 pumping flow rates during low groundwater levels (1.2 to $1.7 \text{ m}^3/\text{s}$) currently exceed the
142 pumping flow rates during high groundwater levels ($0.9 \text{ m}^3/\text{s}$). The minimum groundwater
143 level is fixed at 35 m a.s.l in the main conduit. The maximum drawdown permitted from
144 pumping is thus 30 m below the overflow threshold of the spring (65 m a.s.l., Figure 2b). The
145 lowest water level (i.e. 35 m a.s.l.) was reached during the 1995 hydrological cycle. For
146 environmental reasons, a reserve flow rate of $0.160 \text{ m}^3/\text{s}$ is restored for the Lez River
147 downstream of the spring when it is not overflowing.

148 **2.1.2 Hydrogeological background**

149 The conceptual scheme of the Lez aquifer, built by Salado and Marjolet (1975) and
150 completed recently by Bicalho et al. (2012), shows that the water from the Lez spring
151 comprises a mixture of water from three main units in the aquifer: i) water from the aquifer in
152 the Upper Jurassic limestone and the Lower Cretaceous; ii) surface water (losses) after

153 interacting with the Cretaceous formations; and iii) water from deep circulation in the
154 underlying Middle Jurassic, having long residence time.

155 The hydrogeological functioning of the Lez karst system has been characterized using
156 various rainfall-discharge modelling approaches accounting for pumping (Guilbot, 1975;
157 Thiery and Bérard, 1984; Fleury et al., 2009). These works showed that pumping during low
158 water periods draw out reserves coming from less transmissive zones in addition to the well-
159 drained reserves. An assessment of this pumping influence area around the network of karst
160 conduits give values of about 60 km² (Ladouche et al., 2014), representing only 15% of the
161 Lez spring's catchment.

162 Recent semi-distributed modelling approaches have given a first assessment of the
163 contributions of the main karst units. Simulating the groundwater level in the main drain at
164 the karst outlet using a semi-distributed lumped model, Ladouche et al. (2014) showed that
165 the eastern part of the Matelles fault is contributing 2-fold higher than the western part. This
166 result is coherent with previous works of Kong A Siou (2011) applying neuron models in a
167 semi-distributed approach.

168 **2.2 Hydrogeological data**

169 **2.2.1 Measured data**

170 Hydrogeological measurements have been carried out since 01 June 1974 to 31
171 December 2011. Figure 1 shows the locations of the monitoring sites from which data are
172 used in this study.

173 Daily precipitation intensity was measured in the three weather stations at St-Martin-de-
174 Londres, Valflaunes and Montpellier-Fréjorgues (Météo France, 2012).

175 Pumping discharge rates (Qp) in the karst conduit near the spring were recorded at daily
176 intervals from 1974 to 2000, except during the 1991 to 1996 period when measurements were
177 recorded at weekly intervals. Since 2000, Qp has been recorded at hourly intervals.

178 Until 1982, the piezometric level (h) was measured in the Lez spring basin at daily
179 intervals. Since 1983, h has been measured directly in the main conduit in a borehole located
180 upstream from the spring (Figure 2). Groundwater levels have been recorded on a daily basis
181 from 1974 to 2000, except during the 1991-1996 period when measurements were recorded at
182 weekly intervals. Since 2000, h has been recorded at hourly intervals. Since h expresses
183 groundwater level relative to mean sea level, we also define a drawdown value s which
184 expresses the groundwater level measured from the maximum head h_{max} ($s(t) = h_{max} - h(t)$).
185 The value s was used instead of h in wavelet analysis in order to correlate an increased
186 pumping rate with increased groundwater fluctuation (as the pumping rate increase is
187 inversely correlated with the piezometric level).

188 Measured discharge at the Lez spring is denoted residual discharge (Qr), because the
189 gauging station (Banque Hydro, 2010) is located 300 m downstream the outlet were pumping
190 are carried out (Figure 2). Qr was measured between 1987 and 2007 at daily intervals. Data
191 were corrected from the restored discharge ($0.160 \text{ m}^3/\text{s}$) to the Lez River when the spring was
192 dry. Before 1987, Qr was estimated from water level measurements in the basin. Minimum
193 Qr value is zero.

194 **2.2.2 Calculated data**

195 Precipitation data (P) used in this study is the rainfall time series calculated by
196 Ladouche et al. (2014) to optimize the contribution of three rain gauges in their developed
197 transfer model used to simulate the spring discharge. The method - given by Pinault and

198 Allier (2007) - requires to compute the weighting factor of each rain gauge in order to
 199 maximize the cross-correlation (see Section 3.1 for equations) between P and residual
 200 discharge (Q_r) measured between 1987 and 2007. The linear combination obtained for P is:

$$201 \quad (\text{Eq. 1}) \quad P = 0.33P_1 + 0.54P_2 + 0.13P_3$$

202 where P_1 is the precipitation at St-Martin-de-Londres, P_2 is the precipitation at Valflaunes,
 203 and P_3 is the precipitation at Montpellier-Fréjorgues (Figure 1).

204 During high water period - when overflows are observed at the spring (residual
 205 discharge $Q_r > 0$) - pumping is inferior to the natural discharge (Q_n). Q_r is interpreted as the
 206 difference between Q_n and Q_p : $Q_r = Q_n - Q_p$. According to Ladouche et al. (2014), pumping
 207 mobilizes water stored only in the large conduit (where pumps are localised) during this
 208 period, because no drawdown related to pumping was observed on piezometric $h(t)$ time
 209 series, as well as in the more distant connected piezometer Claret well (see Fig. 1 for
 210 location). Consequently, during high water period, we assume that the storage flow mobilized
 211 by pumping (Q_s) is negligible (almost equal to 0).

212 During low water period - when pumping has dried the spring ($Q_r = 0$) - pumping is
 213 superior to Q_n , and Q_p is interpreted as the sum of Q_n and the Q_s (Ladouche et al., 2014): Q_p
 214 $= Q_n + Q_s$. Pumping thus mobilizes water reserves in the karst system that are inaccessible or
 215 almost inaccessible in natural conditions. This phenomenon is reflected in a decrease of the
 216 piezometric level in the karst conduit (groundwater level < 65 m a.s.l., Figure 2b).

217 The storage flow mobilized by pumping Q_s is calculated by equation 2 (Ladouche et al.,
 218 2014):

$$219 \quad (\text{Eq. 2}) \quad \begin{array}{l} \text{if } Q_p \geq Q_n \text{ then } Q_s = Q_p - Q_n \text{ and } Q_r = 0 \\ \text{if } Q_p < Q_n \text{ then } Q_s \approx 0 \text{ and } Q_r = Q_n - Q_p \end{array}$$

220 This last equation requires to assess Q_n which is unknown during low water level
221 periods, as the Lez spring has been used since 1854 to supply drinking water (Paloc, 1979).
222 During high water level periods (Figure 2a), Q_n is higher than the pumping rate (Q_p), and can
223 be calculated as follows: $Q_n = Q_r + Q_p$. During periods of low water levels (Figure 2b), Q_n
224 cannot be estimated from measurements. Recently, Ladouche et al (2014) have simulated the
225 Q_n time series using a transfer model combining a fast and a slow impulse response. Impulse
226 responses were calculated by inverse modelling during high flow periods when $Q_n = Q_r +$
227 Q_p . The whole Q_n time series was then simulated from 1974 to 2011 using the Tempo
228 software (Pinault, 2001 ; Pinault et al., 2001a, b).

229 All hydrogeological time series for rainfall (P), Lez discharge (Q_r and Q_n), pumping
230 discharge (Q_p) and piezometric levels (h) for the period from 01 June 1974 to 31 December
231 2011 were synchronized at a time interval of 1 day.

232 **2.2.3 Hydrogeological variables used to assess the karst response to** 233 **stressors**

234 This section presents the hydrogeological variables used as input and output to
235 characterize the response of the karst system to climatic and anthropic stressors. Table 1
236 provides an overview of the set of variables, also showing the framework for interpreting their
237 possible inter-relationships. The purpose of this guide is to help to interpret the results of the
238 wavelet analysis used in this article.

239 The input variables representing specifically climatic pressure and anthropic pressures
240 are precipitation P and pumping Q_p , respectively. As the Q_s variable is defined from Q_n and
241 Q_p , (Eq. 2), and Q_n is highly correlated to P , Q_s integrates these both climatic and anthropic
242 pressures. With regard to the output variables, the residual discharge Q_r qualifies the karst

243 response to the rainfall during high flows periods. The water level (h) and especially the
244 drawdown (s) qualify the karst stored changes due to inputs solicitation (precipitation and
245 pumping).

246 Two ambiguous input-output relationships for the defined set of variables were
247 identified. The P - h relationship is disrupted by the impact of pumping on the water level h
248 (major drawdown during the summer). Similarly, Qp cannot be used to study properly the
249 impact of pumping on Qr , since the latest is controlled by both P and Qp variables.
250 Consequently, these two relationships were removed from the presented analysis. We will
251 focus our analysis on the P - Qr relationship to assess the impact of the climate pressure on the
252 karst functioning. The Qp - s relationship is used to assess the anthropic impact on the
253 reserves. Finally, Qs - Qr and Qs - s relationships are used to assess the vulnerability of the
254 resource to both climatic and anthropic pressures.

255 **3 Wavelet analysis**

256 The functions used are briefly presented on the basis of definitions put forward by
257 several authors for wavelet analysis in geosciences (Kumar and Foufoula-Georgiou, 1997;
258 Torrence and Compo, 1998; Labat et al., 2000; Bayazit and Aksoy, 2001; Grinsted et al.,
259 2004; Jevrejeva et al., 2003; Maraun and Kurths, 2004). Wavelet transform can be used to
260 decompose a time series over a time-scale space, thus providing a visualization of power
261 distribution along time and frequency. It is suitable for analysis of non-stationary processes
262 that contain multi-scale features, detection of singularities, or transient phenomena (see the
263 review of Kumar and Foufoula-Georgiou, 1997). Thus, wavelet analysis gives a time-scale
264 representation of the processes and of their relationships.

265 **3.1 Continuous wavelet transforms**

266 The wavelet transform can be seen as a bandpass filter of uniform shape and varying
267 location and width (Torrence and Compo, 1998). The continuous wavelet transform (CWT)
268 $W_x(\tau, a)$ of a time series $x(t)$ is given as follows:

269 (Eq. 3)
$$W_x(\tau, a) = \int_{-\infty}^{+\infty} x(t) \Psi_{\tau,a}^*(t) dt$$

270 (Eq. 4) where
$$\Psi_{\tau,a} = \frac{1}{\sqrt{a}} \Psi\left(\frac{t-\tau}{a}\right)$$

271 represents a group of wavelet functions, $\Psi_{\tau,a}$, based on a mother wavelet Ψ which can be
272 scaled and translated, modifying the scale parameter a and the translation parameter τ
273 respectively. $\Psi_{\tau,a}^*$ corresponds to the complex conjugate of $\Psi_{\tau,a}$. Wavelet functions have multi-
274 scale properties, dilating or contracting a ($a>1$; $a<1$). When a increases, the wavelet covers a
275 higher signal window. It allows the large-scale behaviour of x to be extracted. Conversely,
276 when a decreases, the analysed signal window decreases, allowing local variations of x to be
277 studied. Wavelet transform is thus characterized on the space scale by a window decreasing in
278 width when we focus on local scale structures (high frequency), and widening when we focus
279 on large scale structures (low frequency).

280 As in the Fourier analysis, a wavelet power spectrum (WPS) (also called a scalogram)
281 $P_x(\tau, a)$ can be defined as the wavelet transform of $W_x(\tau, a)$:

282 (Eq. 5)
$$P_x(\tau, a) = (W_x(\tau, a) W_x^*(\tau, a)) = |W_x(\tau, a)|^2$$

283 The choice of the appropriate analysis wavelet depends on the nature of the signal and
284 on the type of information to be extracted from the time series (De Moortel et al., 2004). In
285 this paper, we use the Morlet wavelet, as it is fairly well localized in both time and frequency
286 space (Torrence and Compo, 1998). Other wavelet basis functions, such as Paul and Mexican

hat (DOG), were also tested in order to obtain better time localization, but gave the fairest results in both cases. Statistical significance level was estimated against a red noise model (Torrence and Compo, 1998, Grinsted et al., 2004). As CWTs are applied to time series of finite length, edge effects may appear on the scalogram, leading to the definition of a cone of influence (COI) as the region where such effects are relevant (Torrence and compo, 1998). The COI is marked as a shadow in the scalogram.

The covariance of two time series x and y is estimated using a cross wavelet spectrum (XWT) (also called a cross scalogram) $W_{xy}(\tau, a)$, which is the convolution of the scalogram of both signals:

$$(Eq. 6) \quad W_{xy}(\tau, a) = \left(W_x(\tau, a) W_y^*(\tau, a) \right)$$

XWT reveals an area with a high common power value, but Maraun and Kurths (2004) reported that it appears unsuitable for significance testing of the interrelation between two series. These authors recommend the use of wavelet coherence (WTC) which is a measure of the intensity of covariance of the two series in the time-scale space. Beginning with the approach of Torrence and Webster (1999), the WTC of two time series x and y is defined as:

$$(Eq. 7) \quad C_{xy}^2(\tau, a) = \frac{|S(a^{-1}W_{xy}(\tau, a))|^2}{S(a^{-1}|W_x(\tau, a)|^2) \cdot S(a^{-1}|W_y(\tau, a)|^2)}$$

where S is a smoothing operator in both time and scale (see Torrence and Webster (1999) and Jevrejeva et al. (2003) for detailed mathematical expressions). The 5% significance level of WTC against red AR1 noise is estimated using Monte Carlo methods (Grinsted et al., 2004).

Neighbouring scales and times contain redundance information and are correlated (Maraun and Kurths, 2004), since the wavelet is translated continuously. However, some

approaches exist to limit this redundancy, as is the case with Discrete wavelet analysis (Labat et al., 2000; Bayazit and Aksoy, 2001) based on a wavelet with an orthogonal form.

3.2 Discrete wavelet transform and multiresolution analysis

In order to implement the wavelet transform on sampled signals, the discrete wavelet transform (DWT) can be used to discretize the scale and location parameters j and k , respectively. The discrete form of the wavelet transform of a time series $x(t)$ is given according to Eq. 8:

$$(Eq. 8) \quad W_x(\tau_0, a_0) = \sum_{-\infty}^{+\infty} x(t) \Psi_{\tau_0, a_0}^*(t) dt$$

$$(Eq. 9) \text{ where } \Psi_{\tau_0, a_0} = \frac{1}{\sqrt{a_0^j}} \Psi\left(\frac{t - k a_0^j \tau_0}{a_0^j}\right)$$

with a_0^j being the scale parameter, τ_0 the translation parameter, k and j integers. Ψ_{τ_0, a_0}^* corresponds to the complex conjugate of Ψ_{τ_0, a_0} .

Multiresolution analysis (MRA) is able to study of signals represented at different resolutions. It can be used to decompose a signal into a progression of successive approximations and details in increasing order of resolution. Choosing particular values of a_0 and τ_0 , in Eq. 8, namely $a_0 = 2$, and $\tau_0 = 1$, corresponds to the dyadic case used in MRA. The aim is to reduce/increase the resolution by a factor of 2 between two scales. Therefore, the approximation of a signal $x(t)$ at a resolution j , denoted by A_x^j , and the detail of the same function at a resolution j , denoted by D_x^j , are defined by:

$$(Eq. 10) \quad A_x^j(t) = \sum_{k=-\infty}^{+\infty} C_{j,k} \Psi_{j,k}(t)$$

$$(Eq. 11) \quad D_x^j(t) = \sum_{k=-\infty}^{+\infty} D_{j,k} \Phi_{j,k}(t)$$

328 where $\Phi_{j,k}(t)$ is a scaled and translated basis function called the scaling function, which
 329 is determined with $\Psi_{j,k}(t)$ when a wavelet is selected. $C_{j,k}$ is the scaling coefficient given the
 330 discrete sampled values of $x(t)$ at resolution j and location k . It is calculated from $\Phi_{j,k}(t)$ in a
 331 similar way for the wavelet coefficient $D_{j,k}$ from $\Psi_{j,k}(t)$ (see Kumar and Foufoula-Georgiou
 332 (1997) for detailed mathematical expressions).

333 The signal $x(t)$ can be reconstructed from the approximation and detail components as:

334 (Eq. 12)
$$x(t) = A_x^J(t) + \sum_{j=1}^J D_x^j(t)$$

335 where J is the highest resolution level considered. Since MRA ensures variance is well
 336 captured in a limited number of resolution levels, analysis of energy distribution in the
 337 sampling time series across scales give a good idea of the energy distribution across
 338 frequencies.

339 The choice of wavelet may influence the decomposition, particularly in low frequencies
 340 (Kumar and Foufoula-Georgiou, 1997). We accordingly tested various wavelet functions
 341 (Haar, Battle, Beylkin, Coiflet, Daubechies, Symmlet, Vaidyanathan) in order to assess the
 342 dispersion of results. Since the results were similar overall in the high frequency domain and
 343 less influenced in the lowest, we opted for the frequently used Daubechies 20 wavelet.

344 In order to quantify the relationship quality between two signals across scales, we used
 345 a multiresolution cross-analysis, combining multiresolution with cross-correlation (Labat et
 346 al., 2002). Cross-correlation can be used to determine the degree of similarity between two
 347 signals or two components (at the same resolution level for instance). The cross-correlation
 348 function (CCF) R_{xy} of two time series x and y , is calculated as follows:

349 (Eq. 13)
$$R_{xy}(m) = \frac{c_{xy}(m)}{\sigma_x \sigma_y}$$

350 (Eq. 14) with
$$C_{xy}(m) = \begin{cases} \frac{1}{n} \sum_{t=1}^{n-m} (x_t - \bar{x}) (y_{t+m} - \bar{y}) & \text{for } m \geq 0 \\ \frac{1}{n} \sum_{t=1}^{n+m} (y_t - \bar{y}) (x_{t+m} - \bar{x}) & \text{for } m < 0 \end{cases}$$

351 where C_{xy} is the cross-correlogram, m is the time lag, n is the length, and $\bar{x}$, and σ_x , and
 352 $\bar{y}$ and σ_y , are the average and the standard deviation of x and y , respectively.

353

354 More treatments of the wavelet transform (both continuous and discrete) and wavelet-
 355 based multiresolution (multi-scaling) analysis can be found in Chui (1992), Kumar and
 356 Foufoula-Georgiou (1997), and Mallat (2009) to which the reader is referred for more detail.
 357 Continuous wavelet analyses (CWT, XWT, and WTC) were carried out using a free Matlab
 358 software package (Mathworks, Natick, MA) kindly provided by Grinsted et al. (2004) at
 359 <http://noc.ac.uk/using-science/crosswavelet-wavelet-coherence>. The package includes code
 360 originally written by C. Torrence and G. Compo, available at:
 361 <http://paos.colorado.edu/research/wavelets/>, and by E. Breitenberger of the University of
 362 Alaska, adapted from the freeware SSA-MTM Toolkit: <http://www.atmos.ucla.edu/tcd/ssa/>.
 363 Multiresolution analysis was carried out using a free Matlab software package provided by
 364 the WaveLab Development Team and available at <http://statweb.stanford.edu/~wavelab/>.

365 **4 Results**

366 **4.1 Variability of hydrogeological time series**

367 This Section aims to describe the main variations over hydrological cycles from daily to
 368 annual data, in order to help readers to interpret the results of the wavelet analysis given in the
 369 next Sections 4.2 and 4.3.

370 4.1.1 Daily variability

371 Figure 3a shows daily hydrogeological time series of the Lez karst system over the last
372 38 years. Overall, the Mediterranean climate has a high rainfall intensity, for example
373 attaining 177.5 mm/day in autumn 1976. The mean pumping discharge is around 1 to 1.1
374 m³/s, exceeding the natural baseflow of the Lez spring. To better describe the variability of
375 other hydrogeological data within a given year, Figure 3b zooms in on the 2002-2003
376 hydrological year. Heavy rainfall occurs in the end of the summer and in autumn with an
377 intensity of up to 140 mm/day, generating the highest annual peak flow (15.4 m³/s for Q_r) at
378 the Lez spring and a rise in groundwater level (h) from 65 to 69 m a.s.l.. In the winter and
379 spring seasons, lower rainfall intensities generate lowest hydrodynamic response of the karst
380 system as shown by small flood peaks inferior to 5 m³/s and low groundwater variations.
381 During this period of high groundwater level, Q_p is near to 1.0 m³/s and consistently below
382 Q_n : the spring is discharging ($Q_r > 0$). At the beginning of the dry season in summer, Q_p
383 increases to 1.4 m³/s and exceeds Q_n . The spring consequently dries up (Q_r is zero), and h
384 starts to decrease (water is pumped directly from the conduit). During the low groundwater
385 period in summer, the drawdown reaches 25 m (i.e. the piezometric level drops to 40 m a.s.l.).
386 The saturated zone is highly mobilized by pumping and the reserve storage flow (Q_s)
387 increases to high values (around 1.0 m³/s). The first abundant autumn rainfall in September
388 2003 led to rapid groundwater recovery to the initial level of 65 m a.s.l. as observed on the
389 groundwater level time series (h). This phenomenon occurred in less than one day. The spring
390 then starts to discharge again ($Q_r > 0$) and another similar hydrological cycle begins.

4.1.2 Annual variability

To identify humid and dry periods and changes in the pumping discharge on an annual time scale, Figure 4 shows the annual hydrogeological time series expressed as deviation from the mean. The highest peaks for precipitation P occur in 1976-78, 1987-88, 1995-97, 2000-01, 2002-04 and 2008-09. Although annual discharge peaks (Q_r) are clearly related to these humid years, there is no evidence of any relationship with groundwater level (h). Regarding annual pumping (Q_p) and karst storage flow mobilized by pumping (Q_s), we observed two periods. From 1974 to 1984 both time series are consistently below the mean (except in 1979-80 for pumping), unlike the period from 1984 to the present, during which the time series are higher for most of the time. These periods correspond respectively to a high and a low annual mean groundwater level, meaning that the groundwater level is primarily linked to pumping, leading to a rise in Q_s as pumping rates increase. These results highlight the changes in water resource management from 1983, when water abstraction from deep boreholes began. This allows a higher pumping rate, less constrained in terms of drawdown. Before 1983, the position of the pump in the basin of the spring did not allow the water level to drop to more than 7 m, as opposed to 30 m today.

4.2 Multiresolution analysis

4.2.1 Energy distribution across scales

The main aim of this section is to visualize the distribution of energy across scales (or resolution levels) of the hydrogeological time series. Multiresolution was performed on daily data and the results for the first 10 multiresolution levels are shown in Figure 5. Overall, energy is distributed variably across levels in the hydrogeological time series. Regarding

413 input signals, precipitation P showed high energy mainly for levels 1 to 4 in the high
 414 frequency domain (corresponding to 1 to 8 days). This means that rainfall events on several
 415 days explain most of the variance in the overall P signal. The pumping discharge rate Qp
 416 showed high energy at all levels. Daily to weekly Qp variations are clearly highlighted as
 417 noise in the first levels. For levels 6 to 10, we observed a gradual decrease in energy during
 418 the 1980s. This was clearly visible on raw data (see Figure 3a). As with the pumping signal,
 419 the karst storage flow (Qs) showed high energy across levels. On the other hand, it shows a
 420 gradual energy increase in the 1980s. Regarding the output data for the karst system, the both
 421 residual discharge (Qr) and groundwater level (h) showed high energy distribution across
 422 scales, but different fluctuations over time. For Qr , energy distribution appears to be related to
 423 flood events for high frequencies (levels 1 to 4, corresponding to 1 to 8 days), and to seasonal
 424 and annual variations for lowest frequencies (levels 8, corresponding to 128 days). Energy
 425 variations for the groundwater level are in the same range regardless of the resolution level,
 426 meaning that scale has no apparent effect on groundwater variance. For all levels, h energy
 427 fluctuations over time in the 1980s are consistent with the previously observed gradual
 428 increase and decrease of Qs and Qp respectively.

429 In order to quantify the energy by multiresolution levels, the standard deviation (s.d.)
 430 was calculated by levels for each standardized time series (Figure 6). Overall, as previously
 431 described, P energy distribution across levels is totally different from other hydrogeological
 432 data. Figure 6 shows that s.d. of P decreases from 0.53 to about 0 for low to high
 433 multiresolution levels, meaning that the higher the frequency domain, the higher the energy.
 434 For other time series, we observe a similar main s.d. peak at level 8 (128 days), meaning that
 435 the highest energy is observed for medium levels, corresponding to intra-annual (seasonal)

436 periods. A second, lower s.d. peak is observed at levels 10 (512) and 12 (5.6 years) showing
437 that high energy is also observed for annual and multi-year periodicities. Nevertheless, the
438 s.d. of these time series is not negligible in the first levels, meaning that unlike precipitation,
439 energy is still important across scales.

440 **4.2.2 Multiresolution cross-correlation**

441 To identify the frequency domain where input signals may influence karst system
442 behaviour, in this section we present a cross-correlation function (CCF) across multiresolution
443 levels (Figure 7). As shown in Table 1, the input signals used are P , Qp , and Qs and the
444 output signals used are Qr , and s (instead of h). For each plot, two types of CCF were carried
445 out. In a first case, as proposed by *Labat et al.* (2002), a CCF was carried out between two
446 signals at the same multiresolution level j (black circles). In a second case, we chose to carry
447 out a CCF between an overall input signal (i.e. a non-decomposed time series) and an isolated
448 output signal at a given multiresolution level j (green stars). The maximum cross-correlation
449 values R_{max} are shown in Figure 7 as a function of the multiresolution levels of the output
450 signal expressed in days (at level j , the resolution corresponds to 2^{j-1} days).

451 In the first case (CCF between two signals at the same multiresolution level), the higher
452 the multiresolution level, the higher the value of R_{max} up to 1. In contrast, we observed a R_{max}
453 peak in the second case (CCF using an overall input signal). This difference shows that the
454 output signal at a given multiresolution level is strongly influenced by the energy at lower
455 resolution levels of the input signal. We can thus hypothesize that in the first case the CCF
456 was controlled mainly by the resolution level of the decomposed times series. The R_{max} values
457 of 1 (indicating that the output signal is exactly the same as the input signal) for the highest
458 multiresolution levels are compatible with this hypothesis. Consequently, only the CCFs

459 between an overall input signal and an isolated output signal at a given multiresolution level
460 (green stars) were considered for the analysis of multiresolution cross-correlation.

461 Regarding CCF between precipitation values as input and residual discharge values as
462 output ($P-Q_r$ plots in Figure 7a), we observe a very similar evolution of R_{max} across scales.
463 The highest correlation (around 0.30) occurs for levels 2 and 3, corresponding to 2 and 4-day
464 resolution periods. This means that time series mainly co-vary in the high frequency domain
465 on the flood event-time scale. A sill for levels 7 and 8 (64 and 128 days) is observed, leading
466 the curve to decrease irregularly. This means that highest flood events imprint the discharge at
467 the seasonal scale. For the highest multiresolution levels, both the rainfall and discharge time
468 series become uncorrelated.

469 Regarding a CCF between pumping as input and drawdown as outputs (Q_p-s plots in
470 Figure 7b), we observe a bimodal distribution of R_{max} across levels at level 8 (128 days) and
471 level 13 (11.2 years) with R_{max} of 0.47 and 0.40, respectively. These results show that time
472 series co-vary mainly for medium (intra-annual) and high (multi-year) levels, and that data are
473 uncorrelated in the high frequency domain.

474 Regarding a CCF between storage flows as input and residual discharge and drawdown
475 as outputs (Q_s-Q_r and Q_s-s plots in Figure 7c and 7d, respectively), we observe a similar
476 bimodal distribution of R_{max} across scales, compared to the CCF using Q_p as input. At level 8,
477 however, the highest correlations using Q_s as input (0.39 and 0.61 for Q_s-Q_r and Q_s-s
478 respectively) showed that Q_s is a better signal than Q_p for characterizing the karst response to
479 anthropic pressure within a given year. This is especially true for s as an output signal,
480 because a strong correlation is also observed for the whole spectrum of energy as evidenced

481 by the R_{max} (0.84) of the CCF using an overall signal for input and output (dotted blue line in
482 Figure 7d).

483 **4.3 Continuous wavelet analysis**

484 **4.3.1 Wavelet power spectrum**

485 Multiresolution cross-correlations provide information on frequency domains which are
486 or are not correlated between two signals, but give no information on the temporal variability
487 of their inter-relationships. The aim of this section is to investigate the short- to long-term
488 influence of climatic and anthropic pressures on the karst response using Morlet continuous
489 wavelet analysis (CWT). Because multiresolution analysis showed that pumping rate seems to
490 not influence the hydrogeological response in very high frequencies (several days), CWT was
491 carried out in the monthly to multi-annual frequency domain, in which anthropic impact may
492 be investigated. Figure 8 presents scalograms for all hydrogeological variables to assess the
493 spectral power variance of each hydrogeological signal at each level and at each time lag. On
494 scalograms, the x- and y-axes represent the time-scale space, with frequencies expressed as
495 periods in days (high frequencies or low periods at the top of the plot). The z-axis represents
496 the value of the wavelet coefficient with low to high powers in blue to red colors.

497 Regarding the CWT for rainfall (P in Figure 8a), we identify structures in the high
498 frequency domain (less than 128 days) which are not particularly less marked in the case of
499 spring discharge (Qr in Figure 8d). This low signal attenuation in the 32 to 128-day band in
500 Qr highlights a high transmissive function of the infiltration zone of the karst system.
501 Generally, smallest semi-annual structures (128 to 256 days) appear concomitantly with the
502 main annual ones during wet hydrological cycles in both P and Qr signals (i.e. in 1976-78,

1987-88, 1996-98, 2000-01, 2002-04 and 2008-09, see Figure 4). This shows the imprint of rainfall fluctuations on the karst system in autumn and in a lower manner in spring season. A scale-dependent structure for a 100 to 500-day period is observed in 1995 for P , highlighting the multi-scale distribution of energy among the highest rainfall events. A 10 to 8-year component is observed from 1991 to 2003 for P , reflecting a clear variation in large-scale rainfall distribution. This component is visible in the scalograms of Qr , but power is not above the 5% significance level except in the cone of influence (COI). Globally, on the contrary to high and medium frequencies, these results highlight an attenuation of the lowest frequencies in rainfall by the karst system. This is coherent with the multiresolution cross-correlation analysis presented above, showing a poor correlation between rainfall and discharge in the lowest frequencies.

Regarding CWTs for pumping, karst storage flow and groundwater level (Qp , Qs , and h in Figure 8b, 8c, and 8e, respectively), we identified a high variability of periodic structures over time before and after 1985. In the three scalograms, components are visible in the high frequency domain (below 128 days), for a seasonal period (128-256 days) and for an annual period. However, these structures, visible for Qp from 1974 to 1985, disappear from 1985 onwards, except for some small-scale and erratic annual structures. At the same time, all these structures appear for Qs and h in 1985 after a period without any visible component. This result is consistent with the multiresolution analysis showing a gradual decrease in Qp energy on all scales before 1985, and a concomitant gradual increase in Qs and h energies after this date. Again, the change in energy distribution observed since 1985 is related to the change in the water resource operating strategy since the creation of deep boreholes in 1983. On the

525 contrary to rainfall, no significant long-scale structures are visible in these three scalograms,
526 despite high power in low frequencies of Qp and s .

527 **4.3.2 Cross wavelet and coherence analysis**

528 Inter-relationships between signals are investigated using cross wavelet transform
529 (XWT) and wavelet coherence (WTC). Figure 9 presents cross-scalograms between rainfall,
530 pumping and karst storage flow (as input signals) and spring discharge and groundwater level
531 (as output signals) to help characterize the response of the system.

532 **4.3.2.1 Response to climatic variations**

533 Responses to climatic variations were investigated for XWT and WTC between
534 precipitation as input and residual discharges as outputs (P - Qr plots in Figure 9a and 9b).
535 Except for the driest years (notably 1990-1993) coinciding with low wavelet power,
536 significant coherence between P and Qr appears throughout the time-scale space, suggesting
537 strong relationships between both time series at all scales. XWT highlighted an irregular
538 annual component during wet years when seasonal structures are also visible. Common
539 features from the CWT stand out as being significant on the 8 to 10-year band from 1978 to
540 2005, showing a strong link between P and Qr .

541 **4.3.2.2 Response to anthropic variations**

542 Responses to anthropic pressure were investigated for XWT and WTC between
543 pumping as input and drawdown as outputs (Qp - s plots in Figure 9c and 9d). The Qp - s
544 scalogram showed an irregular annual component during mean and dry years when the
545 surplus is not recharged, emphasizing the impact of pumping on the karst aquifer. Seasonal
546 and high frequencies are also visible irrespective of the hydrological cycle (dry or wet)

547 showing that Qp - s relationships within a given year are not dependent on the annual recharge
548 rate. No long-term influence is visible on the Qp - s scalogram.

549 **4.3.2.3 Response to the cumulative effect of climatic and anthropic variations**

550 Responses to the cumulative effect of climatic and anthropic variations were
551 investigated for XWT and WTC between karst storage flow as input and residual discharges
552 and drawdown as outputs (Qs - Qr plots in Figure 9e and 9f, and Qs - s plots in Figure 9g and
553 9h). For both the Qs - Qr and Qs - s cross-scalograms, two main components are clearly visible
554 at 6 months and 1 year from 1985, showing a high level of co-variance between abstraction
555 from groundwater storage and residual discharge and drawdown from the implantation year of
556 the pumps directly in the karstic drain. For XWT Qs - Qr , the 6-month component appears
557 more irregular since it is only visible when the karst system dries rapidly during the spring
558 when the winter and spring recharge is insufficient to maintain a baseflow discharge above
559 the pumping rate ($Qs > 0$ when $Qn < Qp$). For XWT Qs - s , we observed significant coherence
560 throughout the time-scale space, showing strong relationships between both time series at all
561 scales. Here, it is interesting to note that no significant common features in the wavelet power
562 is visible in lowest frequencies, despite high coherence.

563 **5 Discussion**

564 The aim of this study was to identify using wavelet analysis the respective impacts of
565 climatic and anthropic pressures on the water resource of a karst system under active water
566 management. The first challenge was to identify the contributions of both stressors (namely
567 rainfall and pumping rate, respectively) which concomitantly influence the hydrogeological

response. The second challenge was to track the evolution of phenomena over time and thus to identify the parameters leading to changes in the hydrogeological response of the system.

5.1 Frequency domains in which stressors influence the karst response

Multiresolution analysis (MRA) showed varying energy distribution across levels, meaning that the karst response was highly variable from high to low frequencies. Using multiresolution cross correlation, we determined the frequency domain where output signals (discharge and groundwater level) were most closely correlated to input signals (rainfall and pumping). Our results showed that rainfall and spring discharge are highly correlated in the high frequency domain which reflects the hydrogeological response during flood events of typical highly karstified systems. Pumping and storage flow are correlated to discharge and drawdown in the medium and low frequency domains, illustrating seasonal to multi-year relationships.

5.2 Climatic impact

Continuous wavelet transform (CWT) were used as a complementary approach of MRA to track tendencies and potential evolutions of the hydrogeological response. For rainfall and discharge CWTs, we showed that (i) high rainfall intensities are restored as in the discharge signal during flood events, (ii) an annual component is only visible during wet years in both signals (i.e. when extreme rain events occur), and (iii) the 10-year rainfall component was attenuate in the discharge signal. Despite this attenuation of the lowest frequencies observed in rainfall CWT, cross-wavelet transform (XWT) and coherence (WTC) were useful to provide evidence of long-term climate impacts on spring discharge. This 10-year component

590 should be related to rainfall oscillation in the 5 to 10-year band outlined by Pinault (2012) in
591 the western Mediterranean region. The attenuation of the large-scale rainfall component is
592 clearly different from examples of “non-managed” hydrosystems in the literature (Labat et al.,
593 2000; Chinarro et al., 2011; Hao et al., 2012), which showed that in general karst systems
594 filter less high frequencies (< 2 months) as in our study case, but imprint highly the lowest
595 frequencies due probably to inertial processes generated by the storage volume. Likely
596 explanations of the opposite process at large-scale observed in the Lez karst system is the
597 pumping strategy within a given year that regulates the storage variations (increasing or
598 decreasing the pumping rate according to the available water resource). This interesting result
599 shows that possible long-term impacts of rainfall variability due to climate change on the
600 karst system may be masked by the high pumping rate. However, we showed that XWT and
601 WTC may be a useful approach in order to detect it. The question thus arises of whether
602 pumping can modify the hydrogeological regime of the karst system.

603 **5.3 Anthropic impact**

604 Comparing multiresolution results and those of scalograms, a gradual increase in power
605 distribution was observed for the groundwater level in the 1980s and was concomitant with
606 the gradual increase and decrease in power observed for pumping Qp and storage flow Qs
607 respectively. This evolution is generated by the change in pumping strategy since the creation
608 of deep boreholes in 1983, when pumps were placed directly in the saturated zone of the
609 conduit. We assessed the frequency space of this evolution in terms of cross-scalograms and
610 coherence.

611 The impact of pumping on the hydrogeological response was characterized using $Qp-s$
612 XWTs. An irregular annual component was observed during mean and dry years on the

613 scalogram when there is no extra-recharge. Moreover, no large-scale component was
614 highlighted, meaning that any trend exists on the hydrogeological response to pumping. In
615 fact, regulations impose a maximum drawdown in the Lez aquifer (30 m below the spring),
616 limiting the impact of pumping on storage, even after the increase of pumping rate in 1983.

617 Cross wavelet analysis and coherence showed that Q_s is a better signal than Q_p to
618 explain the piezometric levels, with Q_s reflecting both anthropic and climatic pressures as it
619 displays storage mobilization (i.e. pumping during periods of low groundwater levels). For
620 both Q_s - Q_r and Q_s - s cross-scalograms, two main components are clearly visible at 6 months
621 and 1 year from 1983. Regarding the Q_s - s cross-scalogram, the annual and 6-monthly
622 components appear continuous from 1983 to date, excluding irregularities in the 6-monthly
623 component during wet years. Multi-scale structures are also visible from 2 months to 1 year,
624 illustrating the scale-dependence of both series. This illustrates the anthropic and climatic
625 cumulative effects on the storage level, with continuous annual stress from the starting date of
626 active management. The absence of a large scale component between Q_s and s (and between
627 Q_s and Q_r) can be explained by two phenomena: i) the high recharge rate during autumn in
628 Mediterranean areas, refilling the saturated zone each year, and ii) the pumping restriction
629 imposed by regulations stipulating the maximum drawdown. In this setting, the stress on the
630 groundwater resource does not increase from year to year and the present pumping strategy
631 does not affect the drawdown in the long term, avoiding an over-exploitation of the aquifer.

5.4 Implications on the understanding of the karst hydrogeological response

Multiresolution cross-analysis combined with cross wavelet analysis helps to improve our understanding of hydrogeological processes. Relationships between pumping Q_p and groundwater level s , and between storage flow mobilized by pumping Q_s and s , give informations on the storage evolution over time. Regarding multiresolution levels, we observed that Q_p and Q_s have a mainly mid- to long-term influence on the piezometric level in the conduit. In view of the continuous pumping strategy, we may assume that water stored in the fissured rock matrix is highly mobilized by pumping in the conduit during low water level periods. This is consistent with the results of Ladouche et al. (2014) which showed that the pumping induces mobilization of water in less transmissive units. In fact, conduit/matrix relationships generated by pumping have soon been observed during a 1-month pumping test in the Cent-Fonts karst system (Southern France) by Maréchal et al. (2008). The authors showed that both the fissured matrix (several kilometers away from the pumping well) and the conduit network were affected by the test. Indeed, pumping at the Lez aquifer outlet may influence the upstream part of the karst system over a long distance of about 20 km (Ladouche et al., 2014). The Matelles fault (location shown in Figure 1) is a major drainage axis and the direction of groundwater flow in the Lez aquifer has been ascertained by means of artificial tracer experiments (Marjolet and Salado 1975; Bérard 1983) and by interpretation of monitored water levels measured along this fault (especially at the Claret well - Figure 1) (Karam 1989; Conroux, 2007).

653 **6 Conclusion**

654 The aim of this study was to assess the respective impact of climatic and anthropic
655 pressures on groundwater resources in a Mediterranean karst system under active water
656 management. The main interest in our study was a combination of discrete (multiresolution)
657 and continuous wavelet on 38-year hydrogeological time series recorded at the managed Lez
658 karst aquifer (South France). Our main results showed that water management modifies the
659 hydrogeological response at short and large-time scales. We assume that the reason why
660 large-scale rainfall component do not appear in the spring discharge is that groundwater
661 storage is highly affected by pumping. This result shows that possible long-term impacts of
662 rainfall variability due to climate change may be masked by a high pumping rate. Despite an
663 increase of the pumping rate from the 1980s, the stress on the groundwater resource does not
664 increase from year to year. The current regulation of the hydrogeological conditions by
665 controlling the drawdown – and thus the pumping rate - may be the reason why no long-term
666 anthropic influence was identified. This indicates that the aquifer is currently not over-
667 exploited. Thus, in case necessary, we expect that an increase of the pumping rate is again
668 possible. This study highlights the effectiveness of wavelet analysis in characterizing the
669 response variability of karst systems where the hydrogeological regime is modified by
670 pumping. In order to establish water management scenarios under climatic changes, our
671 approach may be useful to help decompose time series, extracting frequencies in which
672 climatic and anthropic components are mainly localised, before their use in modelling
673 approaches.

674

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816 Table 1: Guide to hydrological variables providing keys to interpret the response of the
817 hydrological system to stressors; relationships underlined in grey are used in the present
818 study.

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Stressors		Karst response		
Input variable used to characterize stressors		Output variable used to characterize the karst response		
		<i>Q_r</i>	<i>h</i> or <i>s</i>	
		Residual discharge	Groundwater level or drawdown	
Climatic pressure	<i>P</i>	<i>Input-Output relationships</i>	<i>P-Q_r</i>	<i>P-h</i> or <i>P-s</i>
	Precipitation	<i>Highlight:</i>	Hydrogeological response	<i>Relationship disrupted by pumping</i>
Anthropic pressure	<i>Q_p</i>	<i>Input-Output relationships</i>	<i>Q_p-Q_r</i>	<i>Q_p-s</i>
	Pumping discharge	<i>Highlight:</i>	<i>Relationship disrupted by precipitation</i>	Storage mobilization
Climatic and anthropic pressures	<i>Q_s</i>	<i>Input-Output relationships</i>	<i>Q_s-Q_r</i>	<i>Q_s-s</i>
	Karst storage flow mobilized by pumping	<i>Highlight:</i>	Groundwater stress due to active water management during low groundwater level periods	Level of stress on the groundwater resource

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822 Figure 1: Hydrogeological map of the Lez karst system.

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825 Figure 2: Cross section of the karst network at the outlet of the Lez aquifer, showing the
 826 location of the spring and of the pumping station; Piezometric levels are also plotted
 827 according to high (2a) and low (2b) flow conditions.

830 Figure 3: Hydrogeological daily time series of the Lez karst system from 01 June 1974 to 31
831 December 2011 (a), and a zoom on the 2002-2003 hydrological cycle (b); from top to bottom:
832 precipitation (P), measured pumping discharge (Q_p) and calculated storage discharge
833 mobilized by pumping (Q_s), measured Lez residual discharge (Q_r) and simulated natural
834 discharge (Q_n), and the measured groundwater level in the conduit (h).

837 Figure 4: Annual evolution of hydrogeological time series expressed as deviation from the
 838 mean; the annual hydrological cycle was delimited from August to July.

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Figure 5: Multiresolution analysis of daily hydrogeological time series using Daubechies 20 wavelets; different components of the decomposition correspond from top to bottom to short-scale to long-time scale processes, with level j corresponding to time scales at 2^{j-1} ; grey rectangles indicate a lack of data for the first levels 1 to 3 (1 to 4-day resolution) during a period of weekly raw data.

848 Figure 6: Standard deviation by multiresolution levels (from level 1 to level 14) of the
 849 standardized hydrogeological signals.

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Figure 7: Cross-correlation functions (CCF) applied at different scales of the multiresolution analysis with: P - Qr CCF (7a), Qp - s CCF (7b), Qs - Qr CCF (7c), and Qs - s (7d); two cases are shown for each plot: i) CCF between two signals at the same multiresolution level j (black circles), and ii) CCF between an overall input signal (i.e. a non-decomposed time series) and an isolated output signal at a given multiresolution level j (green stars). Maximum CCF values (R_{max}) are expressed as a function of the multiresolution levels in days (at level j , the resolution corresponds to 2^{j-1} days). R_{max} values are also plotted for CCF between two overall signals (dashed blue line).

Figure 8: Continuous wavelet power spectra of the hydrogeological time series; the thick black outline designates the 5% significance level against red noise and the cone of influence, where edge effects might distort the picture, is shown in a lighter shade.

867 Figure 9: Cross wavelet (XWT) and wavelet coherence (WTC) spectra between (9a and 9b)
868 rainfall and residual Lez discharge $P-Q_r$, (9c and 9d) pumping and drawdown Q_p-s , (9e and
869 9f) groundwater stress and residual Lez discharge Q_s-Q_r , and (9g and 9h) groundwater stress
870 and drawdown Q_s-s ; The thick black outline designates the 5% significance level against red
871 noise and the cone of influence where edge effects might distort the picture is shown as a
872 lighter shade.

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