

1 We would like to thank Anonymous Referee #1 for their detailed review comments. We
2 found them to be insightful, and, through our responses to them set out below, we believe
3 that they have resulted in a much improved paper.

4 Major Comment 1: Referee #1 states that “it is not correct to say that this paper quantifies
5 the teleconnection contribution to the absolute groundwater variability for the first time (line
6 346, 392, 449, 509). The authors claim that all previous studies performed low-pass filtering
7 or some averaging of groundwater level time-series before wavelet transform or PCA
8 methods. This is not so, at least in the case of Tremblay et al., 2011 and Neves et al., 2019.
9 The proportion of groundwater variability driven by teleconnections in the UK seems indeed
10 much lower than in other parts of the world. Blaming the amplification of low frequencies in
11 other studies (that does not happen) is therefore not valid, and the authors should seek other
12 explanations.”

13 *Response to Major Comment 1: The Reviewer is correct that there may be many other*
14 *contributing explanations in some cases which we have now outlined in Lines 375 - 386.*
15 *However, we do also think that previous studies that have sought to quantify the proportion*
16 *of extra-annual cyclical variability in groundwater level and that may have used pre-*
17 *processing steps that might have altered the strength of extra-annual periodicities within the*
18 *groundwater spectra. A key example is cumulative departure from the mean (CDM) which*
19 *has been undertaken by Neves et al., 2019. While not explicitly designed as a low-pass filter,*
20 *CDM is a process that amplifies low frequency periodicity and suppress higher frequency*
21 *periodicities. This is, for example, exemplified in figure 4 in Neves et al., 2019 where we can*
22 *see little annual variability in rainfall; which we would not expect from a ‘raw’ dataset. As a*
23 *result of this, the strength of extraannual periodicities may be misrepresented when*
24 *compared to the raw groundwater level data. Another example is given in Tremblay et al.,*
25 *2011., while no preprocessing of the data is apparent, periodicities reported have not*
26 *included the strength of seasonality. As such, we cannot tell the actual strength (and*
27 *therefore importance) of the extra-annual periods, as we cannot tell how they compare to*
28 *seasonality (known to be a major component of hydrological processes). As such we believe*
29 *this paper provides an explicit assessment of the percentage of cyclical variability to the*
30 *unaltered groundwater level data spectrum. We have amended the text in the locations*
31 *highlighted by Referee #1 to make this clearer, e.g. Lines 375 - 386, 495 – 502.*

32 Major Comment 2: Referee #1 states that “The results may probably be a consequence of
33 the specific climate and hydrogeologic conditions in the UK, but may also be a consequence
34 of the different methodology used to compute the percentages of variance. Do the authors
35 get the same results using SSA or PCA? One alternative method should be used in order to
36 be sure.”

37 *Response to Major Comment 2: We appreciate why Reviewer #1 has made this observation*
38 *and suggests additional analyses. However, SSA/PCA (which the co-authors applied to*
39 *groundwater level observations in Holman et al.(2009)) requires removal of trends (non-*
40 *stationary) before any meaningful information on principal components can be extracted and*
41 *therefore implies stationarity. In addition, the aim of this paper was to identify specific*
42 *periodicity bands that are shared between groundwater hydrographs, and with SSA/PCA*
43 *there is no guarantee that eigenvectors between datasets will be comparable or even*
44 *periodicities of these can be confidently estimated (as one would have to again assume*
45 *stationarity to identify frequencies from principal components). Nevertheless, we have*
46 *extended our literature review to include potential other sources for these signal strengths in*
47 *light of the Reviewer’s comment. E.g. Lines 479 – 492, 375 – 386*

48 Major Comment 3.1: Referee #1 states that “A closer look at Figure 4 shows time intervals
49 between droughts of approximately 2.5, 3, 5, 6 and seven years. Therefore, it seems
50 excessive to declare that the approach presented in this paper can be used to predict
51 droughts with a recurrence of seven years (line 492).”

52 *Response to Major Comment 3.1: We agree with the comment that the wording around the*
53 *recurrence of drought events is too strong and does not account for the variability in the time*
54 *intervals between recorded droughts. To address this concern, we have now added a further*
55 *review of drought mechanisms and have updated the text to refer to reflect drought risk,*
56 *rather than the definite timings of drought in Lines 450 – 492. In addition, Figure 4 has been*
57 *modified to better illustrate the drought start/end dates, although there is inevitable spatial*
58 *uncertainty in these.*

59 Major Comment 3.2: Referee #1 states “Moreover, the authors do not even mention the non-
60 stationarity of teleconnections and ignore the effects of global warming on the predictability
61 and statistics of extreme events. The authors need to elaborate more on these issues.”

62 *Response to Major Comment 3.2: We agree with the Reviewer that more elaboration is*
63 *needed on these issues, although we also note that the effects of global warming on the*
64 *predictability and statistics of extreme events is a very broad and still developing subject. It is*
65 *mentioned in the text that the varying strength (and therefore the non-stationarity) of the*
66 *NAO does not directly appear to influence the occurrence of historical drought, therefore*
67 *wide-spread droughts appear sensitive to the NAO phase, rather than its overall strength.*
68 *However, we have now also added additional text to clarify these issues in Lines 479 – 492.*

69 Minor Comment 1: Please increase the font size of text and labels in the pictures – *Figures*
70 *have been updated*

71 Minor Comment 2: Line 283: can you explain better why the 7-year cycle has greater
72 significance values in rainfall than in groundwater? *Text has been updated in Lines 287-289*

73 Minor Comment 3: Line 315: do you mean misalignments amongst borehole records? Are
74 there consistent misalignments amongst aquifers? *Text has been updated at line 319*

75 Minor Comment 4: Line 321: figure 6 instead of figure 4? *Text has been updated*

76 Minor Comment 5: Lines 342-354: the whole paragraph is redundant and would better be
77 omitted. *We agree that this paragraph is not required and have removed the text*

78

79 Response to Reviewer #2 comments

80 We would like to thank Anonymous Referee #2 for their detailed review comments. We
81 found them to be insightful, and, through our responses to them set out below, we believe
82 that they have resulted in a much improved paper.

83 Major comment 1: - In general the interpretation of trends by aquifer type is tricky for Oolite
84 and Greensand sites as there are only 2 and 3 observation boreholes. I recommend clearly
85 stating the number of observation boreholes in the introduction (somewhere the introduction
86 between line 110 and 117) and afterwards avoiding (over)interpretation of statistic measures
87 in these two aquifer types (e.g. lines 262, 277-278, 290-292, 325- 326 . . .). Furthermore
88 there is no strong differences between the aquifer types, at least I don't see these e.g. in
89 Figure 6, in my opinion these differences are not shown in your results (line 365 – 369).
90 Consider rephrasing to make a less strong claim.

91 *Response to Major comment 1: We agree that it is difficult to interpret patterns in response*
92 *as a function of aquifer type, particularly for the Oolites and Greensands where there are*
93 *only a couple of observations from each aquifer; that we should avoid over interpreting any*
94 *of the aquifer specific results. Consequently, we have revised the text at L112-116 to*
95 *explicitly state how many observations there are for each aquifer, and have added*
96 *cautionary statements in the appropriate sections of text noting the relatively small sample*
97 *sizes and the consequent difficulties in unambiguously identifying systematic differences in*
98 *responses between the different aquifers, e.g. Lines 265 – 267, and we have avoided group-*
99 *specific interpretation in the discussion for these groups.*

100 Major comment 2: The drought events used for comparison, do not occur in the 7-year
101 cycles that are proposed for potentially predicting groundwater droughts in the UK. These
102 drought events occur in different time intervals

103 *Response to Major comment 2: We agree that the wording around the recurrence of drought*
104 *events was too strong and did not account for the different time intervals between recorded*
105 *droughts. In response we have now included a further review of drought mechanisms and*
106 *have updated the text to refer to reflect drought risk, e.g. Lines 469-566, rather than the*
107 *definite timings of drought. In addition, Figure 4 has been modified to better illustrate the*
108 *drought start/end dates, although there is inevitable spatial uncertainty in these.*

109 Major comment 3: To support teleconnection influences of larger scale climate phenomena
110 you need to further elaborate on this. The claims in the discussion on the relation of NAO
111 and EA to the 7 year and 16-32 year cycles of droughts are very strong considering the
112 results; consider reformulating it

113 *Response to Major comment 3: We have now softened our claims regarding the NAO and*
114 *EA control on groundwater and rainfall in the discussion, and included further literature*
115 *review about the potential causes for these signals, see Lines 427 – 432 and 461- 463, and*
116 *have removed lines 464 – 467.*

117 Major comment 4: Key for the interpretation of section 3.2 is additional information on the
118 drought periods you are referring to (green bands in Figures 4&5). It would be helpful to
119 provide some background on these events (on magnitude and durations), this potentially
120 also helps to improve the discussion on climatic teleconnections.

121 *Response to Major comment 4: We have now included additional information on the drought*
122 *periods in the Discussion at Lines 481-557*

123 Major comment 5: The discussion can be (and should be) considerably shortened by
124 removing the first, very general and summarizing paragraph, also the last parts of the
125 discussion are a little more messy than the rest of the manuscript, please consider re-
126 organizing the discussion a little bit (see also minor comments)

127 *Response to Major comment 5: We agree that this paragraph is not required and have*
128 *removed the text at Lines 347 - 359, and have reworded the final paragraph at Lines 576 –*
129 *579 and 581 - 605.*

130 Major comment 6: In my opinion, the quality of the Figures is not sufficient for publication:
131 please change size of labels, axis labels, legends e.g. in Figures 2, 3, 4 and 5. Add a scale
132 bar to all GB maps (Figure 1, Figure 6 and sup. Figure 1) .

133 *Response to Major comment 6: Figures have been updated to include the suggested*
134 *changes*

135 Major comment 7: Also in the conclusions we find some very strong statements that are in
136 my opinion only partially supplied by your results: line 509 “we quantify, for the first time
137 globally” (as pointed out before this is not the first time, see interactive comments); line 517
138 – 523 “. . . allowing the estimation of future drought. . .” (I would suggest changing this very
139 strong claim accordingly, you show potential control of NAO and EA on groundwater
140 droughts in the UK); line 527-529 “it is clear from our results . . . drought prediction and its
141 management across the North Atlantic region” (inn my opinion you cannot say that from your
142 results, you mostly qualitatively analyse the coinciding timing of drought and climate across
143 the UK); I’d skip Interactive comment line 524 – 527 at it is not very informative;

144 *Response to Major comment 7: We have amended the text throughout the document to*
145 *focus more on the contribution to the existing knowledge base rather than claiming*
146 *anywhere to be the first study to produce such findings. See, for example, Lines 495 – 516.*

147

148 Amendments made to the manuscript

- 149 1. All instances of “extra-annual” have been amended to “multi-annual” for clarity
- 150 2. Information on sample sizes has been included in the description of datasets at Lines
151 112, 114, 115 and 116 in response to Reviewer 2, Major comment 1
- 152 3. Scale bar has been added to Figure 1 at line 121 in response to Reviewer 2 Major
153 comment 6
- 154 4. Text size increased in Figure 2, 3, 4 and 5, and scalebar added to figure 6 in
155 response to Reviewer 2 Major comment 6
- 156 5. Text added to caution against over interpretation of datasets with small sample sizes
157 at Lines 265-267 in response to Reviewer 2, Major comment 1
- 158 6. Explanatory text added to Lines 288 – 289 in response to Reviewer 1 Minor comment
159 2
- 160 7. Lines 319-320 have been amended in response to Reviewer 1 Minor comment 3
- 161 8. Lines 347 – 359 have been removed in response to Reviewer 1 Minor comment 5
162 and Reviewer 2 Major comment 5.
- 163 9. Lines 390 – 403 have been amended according to Reviewer 1 Major comment 1 and
164 2 and Reviewer 2 Major comment 3
- 165 10. Lines 446 – 453 have been amended in response to Reviewer 2 Major comment 7.
- 166 11. Lines 461 – 467 have been amended in response to Reviewer 1 Major comment 1
167 and Reviewer 2 Major comment 7
- 168 12. Lines 475 – 567 have been amended in response to Reviewer 1 Major comment 3.1
169 and Reviewer 2 Major comment 2.
- 170 13. Lines 513-517 have been amended in Response to Reviewer 1 Major comment 3.2
- 171 14. Line 571 and Lines 577-579 has been amended in response to Reviewer 2 Major
172 comment 7
- 173 15. Lines 581 – 567 have been amended in response to Reviewer 1 Major comment 3.1
174 and Reviewer 2 Major comment 2.

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178

1 Manuscript with changes tracked

2

3 **Understanding the potential of climate teleconnections to project future groundwater**
4 **drought**

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10

11

12 **Abstract**

13 Predicting the next major drought is of paramount interest to water managers, globally.

14 Estimating the onset of groundwater drought is of particular importance, as groundwater
15 resources are often assumed to be more resilient when surface water resources begin to fail.

16 A potential source of long-term forecasting is offered by possible periodic controls on
17 groundwater level via teleconnections with oscillatory ocean-atmosphere systems. However,
18 relationships between large-scale climate systems and regional to local-scale rainfall, ET and
19 groundwater are often complex and non-linear so that the influence of long-term climate cycles
20 on groundwater drought remains poorly understood. Furthermore it is currently unknown

21 whether the absolute contribution of multi-annual climate variability to total groundwater
22 storage is significant. This study assesses the extent to which ~~inter-annual~~multi-annual

23 variability in groundwater can be used to indicate the timing of groundwater droughts in the
24 UK. Continuous wavelet transforms show how repeating teleconnection-driven 7-year and 16-

25 32 year cycles in the majority of groundwater sites from all the UK's major aquifers can
26 systematically control the recurrence of groundwater drought; and we provide evidence that

27 these periodic modes are driven by teleconnections. Wavelet reconstructions demonstrate
28 that multi-annual periodicities of the North Atlantic Oscillation, known to drive North Atlantic

29 meteorology, comprise up to 40% of the total groundwater storage variability. Furthermore,
30 the majority of UK recorded droughts in recent history coincide with a minima phase in the 7-

31 year NAO-driven cycles in groundwater level, ~~allowing the estimation of future~~providing insight

32 ~~into~~ drought occurrences on a multi-annual timescale. Long-range groundwater drought
33 forecasts via climate teleconnections present transformational opportunities to drought
34 prediction and its management across the North Atlantic region.

35

36 1. Introduction

37 ~~Inter-annual~~Multi-annual variability detected in hydrometeorological datasets has long been
38 associated with systems of atmospheric-oceanic (climatic) oscillation, such as El Niño
39 Southern Oscillation (ENSO) and the North Atlantic Oscillation (NAO). Such periodic
40 teleconnection signals have been detected in rainfall (Luković et al. 2014), evapotranspiration
41 (Tabari et al. 2014), air temperature (Faust et al. 2016), and river flow (Su et al. 2018; Dixon,
42 et al. 2011); however these periodicities are often weak when compared to the finer-scaled
43 (daily to seasonal) variability that is typical of hydrometeorological processes (Meinke et al.
44 2005). By contrast, groundwater systems are expected to be particularly susceptible to ~~inter-~~
45 ~~annual~~multi-annual teleconnection influence, given their sensitivity to long-term changes in
46 rainfall and evapotranspiration (Bloomfield & Marchant 2013a; Forootan et al. 2018; Van Loon
47 2015; Folland et al. 2015), and their ability to filter fine-scale variability in recharge signals
48 (Dickinson et al. 2014; Velasco et al. 2015; Townley 1995). Consequently, recent studies have
49 focused on the detection of long-term periodic cycles in groundwater levels in Europe (e.g.
50 Holman et al. 2009; Holman et al. (2011); Folland et al. (2015); and Neves et al. (2019)), North
51 America (e.g. Tremblay et al. (2011); Kuss & Gurdak (2014)) and globally (e.g. Wang et al.
52 (2015); Lee & Zhang (2011)), and their relationships with climatic oscillations. An
53 understanding of ~~inter-annual~~multi-annual periodicity strength in groundwater level may
54 provide an improvement in long-lead forecasting of hydrogeological extremes (Rust et al.
55 2018; Meinke et al. 2005; Kingston et al. 2006), in part, by enabling such cyclical behaviour to
56 be projected into the future. This is particularly apparent of groundwater drought, which is
57 known to result from multi-annual moisture deficits (Van Loon 2015; Van Loon et al. 2014;
58 Peters et al. 2006). Therefore, it is critical to quantify the absolute strength of all periodicities

59 within groundwater levels so that the strength of ~~inter-annual~~multi-annual cycles, the influence
60 of teleconnections, and their contribution towards groundwater droughts can be understood.

61 Existing studies into groundwater teleconnections use quantitative methods to detect periodic
62 behaviour in groundwater datasets and often their relationship with time series of climate
63 indices (used to measure the strength and state of climate oscillations). Common quantitative
64 methods range from temporal correlation analysis (Knippertz et al. 2003; Szolgayova et al.
65 2014) to more complex periodicity detection and comparison. These latter methods include
66 Fourier transform, (Nakken, 1999, Pasquini et al. 2006), singular spectrum analysis (SSA)
67 (Kuss & Gurdak 2014; Neves et al. 2019) and wavelet transformations (Fritier et al. 2012;
68 Holman et al. 2011; Tremblay et al. 2011). The wavelet transform (WT) has been shown to be
69 particularly skilful at detecting ~~inter-annual~~multi-annual periodic behaviour in noisy
70 hydrogeological datasets; detecting the influence of the NAO, ENSO and Atlantic Multidecadal
71 Oscillation (AMO) on North American groundwater levels (Kuss & Gurdak 2014; Velasco et
72 al. 2015), and the NAO, East Atlantic pattern (EA) and Scandinavian pattern on European
73 groundwater level variability (Holman et al. 2011; Neves et al. 2019). However, in order to
74 enhance ~~inter-annual~~multi-annual periodicity detection, many studies have used data
75 processing methods that remove or suppress variability at the higher end of the frequency
76 spectrum (e.g. winter or annual averaging or conversion of time series to cumulative
77 departures from mean (Weber & Stewart 2004)). Due to this data modification, it is currently
78 unknown whether the absolute contribution of multi-annual climate variability to total
79 groundwater storage is significant. This limitation makes assessment of systematic linkages
80 between climatic oscillations and groundwater level response problematic (Rust et al. 2018).
81 As a result, the fundamental question of whether ~~inter-annual~~multi-annual teleconnection
82 cycles in groundwater level are sufficiently strong to influence hydrogeological drought
83 remains largely unanswered. Given the potential for improved long-lead forecasting,
84 quantification of ~~inter-annual~~multi-annual variability in groundwater level represents an

85 opportunity to support efficient infrastructure investment, systems of water trading (Rey et al.
86 2018) and robust planning for groundwater drought.

87 The aim of this paper is to assess the extent to which periodic behaviour in groundwater level
88 produced by teleconnections, may be used as an indicator for the timing of groundwater
89 droughts. In doing so, this paper develops and applies an improved method to describe and
90 characterise the absolute strength of periodic behaviour in groundwater level and its drivers
91 (rainfall and evapotranspiration). This aim will be met by addressing the following research
92 objectives:

- 93 1. Characterise dominant intra- and ~~inter-annual~~multi-annual periodicities in groundwater
94 level records across a range of aquifer types
- 95 2. Quantify the absolute strength of these ~~inter-annual~~multi-annual periodic groundwater
96 level oscillations compared to the total variability in groundwater levels
- 97 3. Qualitatively assess evidence for the control of climate teleconnections on identified
98 ~~inter-annual~~multi-annual periods
- 99 4. Assess the extent to which the timing of the ~~inter-annual~~multi-annual periodic
100 groundwater level oscillations align with recorded groundwater droughts

101 These objectives will be implemented on UK hydrogeology records, given the considerable
102 coverage of recorded groundwater level data in time and across the country (Marsh &
103 Hannaford 2008) however the methodologies developed can be applied to any regions.

104

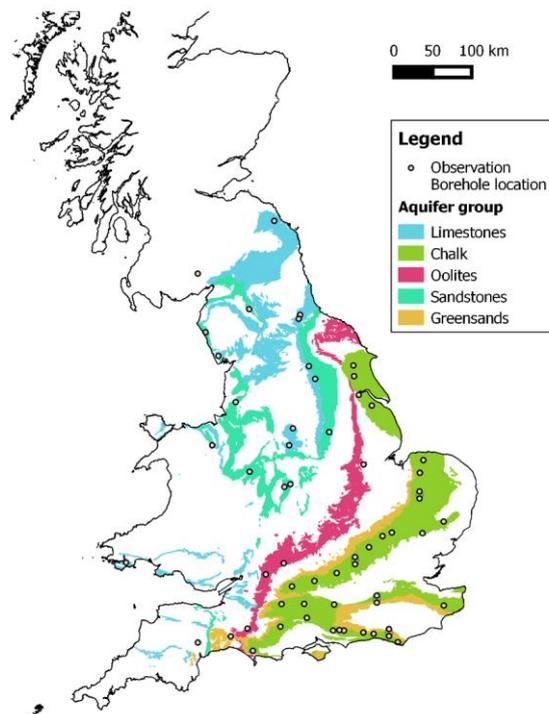
105 **5. Data and Methods**

106 2.1. Groundwater data

107 Groundwater level time series from 59 reference boreholes covering all of the major UK
108 aquifers, with record lengths of more than 20 years and data gaps no longer than 24 months,
109 have been assessed in the study. These recorded groundwater level hydrographs range from
110 21 to 181 years in length, with an average length of 53 years. The sites are part of the British

111 Geological Survey's Index Borehole network and, in addition to their data coverage, have been
112 chosen as they exhibit representative and naturalistic hydrographs with minimal impact from
113 abstractions. They cover a range of unconfined and confined consolidated aquifer types and
114 have been categorised into 5 main aquifer groups; [34 records in](#) -Chalk, a limestone aquifer
115 comprising of a dual porosity system with localized areas where it exhibits confined
116 characteristics; [8 records in](#) Limestone, characterised by fast-responding fracture porosity; [3](#)
117 [records in](#) Oolite characterised by highly fractured lithography with low intergranular
118 permeability; [12 sites in](#) Sandstone, comprised of sands silts and muds with principle inter-
119 granular flow but fracture flow where fractures persist; and [2 records in](#) Greensands,
120 characterised by intergranular flow with lateral fracture flow depending on depth and formation
121 (Marsh & Hannaford 2008).

122



123

124 *Figure 1 - Location of the observation borehole locations used in this study. Boreholes within 0.5 km of another*
 125 *have been displaced and denoted on a grey circle for visibility.*

126

127 2.2. Rainfall

128 Rainfall time series from the Centre of Ecology and Hydrology's CEH-GEAR 1km gridded
 129 rainfall dataset (Tanguy et al. 2016), which is based on spatio-temporal interpolation of daily
 130 rain gauge totals between 1890 and 2017, was used. However, relatively few rainfall stations
 131 exist prior to 1950 that were used for this interpolation; as such data prior to 1950 was not
 132 used in this analysis. Monthly rainfall series have been calculated for each borehole from the
 133 1km grid cell in which they are located, as geospatial data on areas of groundwater recharge
 134 connected to specific observation boreholes does not exist. This dataset may contain artefacts
 135 as a result of the spatio-temporal interpolation, in comparison to station data. However the
 136 use of rainfall data in this study is to provide a broad understanding of rainfall periodicities to

137 supplement those from groundwater level data. As such, this interpolated dataset is deemed
138 appropriate.

139

140 2.3. Potential Evapotranspiration (PET)

141 Monthly PET series for each borehole have been derived from the Centre of Ecology and
142 Hydrology's CHES-PE 1km gridded dataset of calculated daily PET values. The PET values,
143 between 1960 and 2015, were calculated using the Penman-Monteith equation, with
144 meteorological data taken from the CHES gridded meteorological dataset. Details on the
145 underlying observation datasets and interpolation methods can be found in Robinson et al.
146 (2016). This data has been used previously to study long-term trends in hydrological variability
147 (Robinson et al. 2017).

148

149 2.4. Methods

150 2.4.1. Data pre-processing

151 In this study we use the continuous wavelet transform (CWT) to produce a time-averaged
152 frequency spectrum for each borehole hydrograph and co-located rainfall and PET time series.

153 For all datasets, gaps less than two years were infilled using a cubic spline to produce a
154 complete time series for the CWT. This interpolated information was later removed from the
155 time-frequency transformation (prior to time-averaging) to ensure that the data infilling had
156 minimal effect on the final spectrum. For time series with gaps greater than two years, the
157 shortest time period before or after the data gap was removed to produce one complete
158 record. Individual rainfall and PET time series were trimmed to match the length of the
159 corresponding borehole level time series. All time series were centred on the long-term mean
160 and normalized to the standard deviation to produce a time series of anomalies. Unlike most
161 previous studies, no high- or low-band filtering was undertaken on the datasets, ensuring all

162 information on periodic variability was preserved. This approach ensures that the Proportion
163 of a periodicity to the variance (standard deviation) of the original dataset is not modified.

164 2.4.2. Continuous Wavelet Transform.

165 Following the data pre-processing steps, a CWT was applied to quantify the time-averaged
166 frequency spectra of the rainfall, PET and groundwater datasets. The CWT has been used
167 to assess long term trends and periodicities in many hydrological datasets including rainfall
168 (Rashid et al. 2015), river flow (Su et al. 2017), and groundwater (Holman et al. 2011; Kuss
169 & Gurdak 2014). We use the package “WaveletComp” produced by Rosch & Schmidbauer
170 (2018) for all transformations in this paper.

171 The continuous wavelet transform, W , consists of the convolution of the data sequence (x_t)
172 with scaled and shifted versions of a mother wavelet (daughter wavelets):

$$W(\tau, s) = \sum_t x_t \frac{1}{\sqrt{s}} \psi^* \left(\frac{t - \tau}{s} \right) \quad (\text{Eq. 1})$$

173 where the asterisk represents the complex conjugate, τ is the localized time index, s is the
174 daughter wavelet scale and dt is increment of time shifting of the daughter wavelet. The
175 choice of the set of scales s determines the wavelet coverage of the series in its frequency
176 domain. The Morlet wavelet was favoured over other candidates due to its good definition in
177 the frequency domain and its similarity with the signal pattern of the environmental time
178 series used (Tremblay et al. 2011; Holman et al. 2011).

179 The CWT produces a time-frequency wavelet power spectrum for each time series. Within
180 the time-frequency spectra, a cone of influence (COI) is used to denote those parts that are
181 affected by edge-effects, where estimations of spectral power are less accurate. Therefore
182 only data from within COI were averaged over time to produce a time-average wavelet power
183 spectrum for frequency bands from 6 months up to 64 years. Wavelet power spectra were
184 then normalised to the maximum average wavelet value so that the frequency distribution of
185 each site can be directly compared. The normalized average wavelet power spectra (herein

186 referred to as the wavelet power spectra) provide a comparative measure of the strength of
187 the range of periodicities within frequency space.

188 2.4.3. Significance testing

189 As Allen and Smith (1996) demonstrate, geophysical datasets can exhibit pseudo-periodic
190 behaviour as a result of their lag-1 autocorrelation (AR1) properties. Datasets with greater
191 AR1 tend to have spectra biased towards low frequencies, thus they are described as
192 containing red noise (Allen et al. 1996; Meinke et al. 2005; Velasco et al. 2015). In order to
193 assess the likelihood that a periodic signal is the result of internal (red) noise within the data,
194 significance of the red noise null hypothesis was tested. For this, 1000 randomly constructed
195 synthetic series with the same AR1 as the original time series were created using Monte Carlo
196 methods. Wavelet spectra maxima from these represent periodicity strength that can arise
197 from a purely red noise process. Wavelet powers from the original dataset that are greater
198 than these “red” periodicities are therefore considered to be driven by a process other than
199 red noise, thus rejecting the null hypothesis. Here, while a 95% Confidence Interval (CI) (\leq
200 0.05 alpha values) is identified, we report on the full range of alpha results to provide a detailed
201 assessment of the likelihood of external forcing on periodic behaviour.

202 2.4.4. Time reconstruction

203 In order to assess the characteristics of periodicities over time, we employ a reversal of the
204 wavelet transform (wavelet reconstruction) to convert selected periodic domains back into a
205 time series of normalised anomalies. Period bands were selected where the frequency spectra
206 identified shared wavelet power (and significance) between groundwater, rainfall and PET,
207 indicating a wide-spread signal presence at these bands.

208 The reverse wavelet transform is given by:

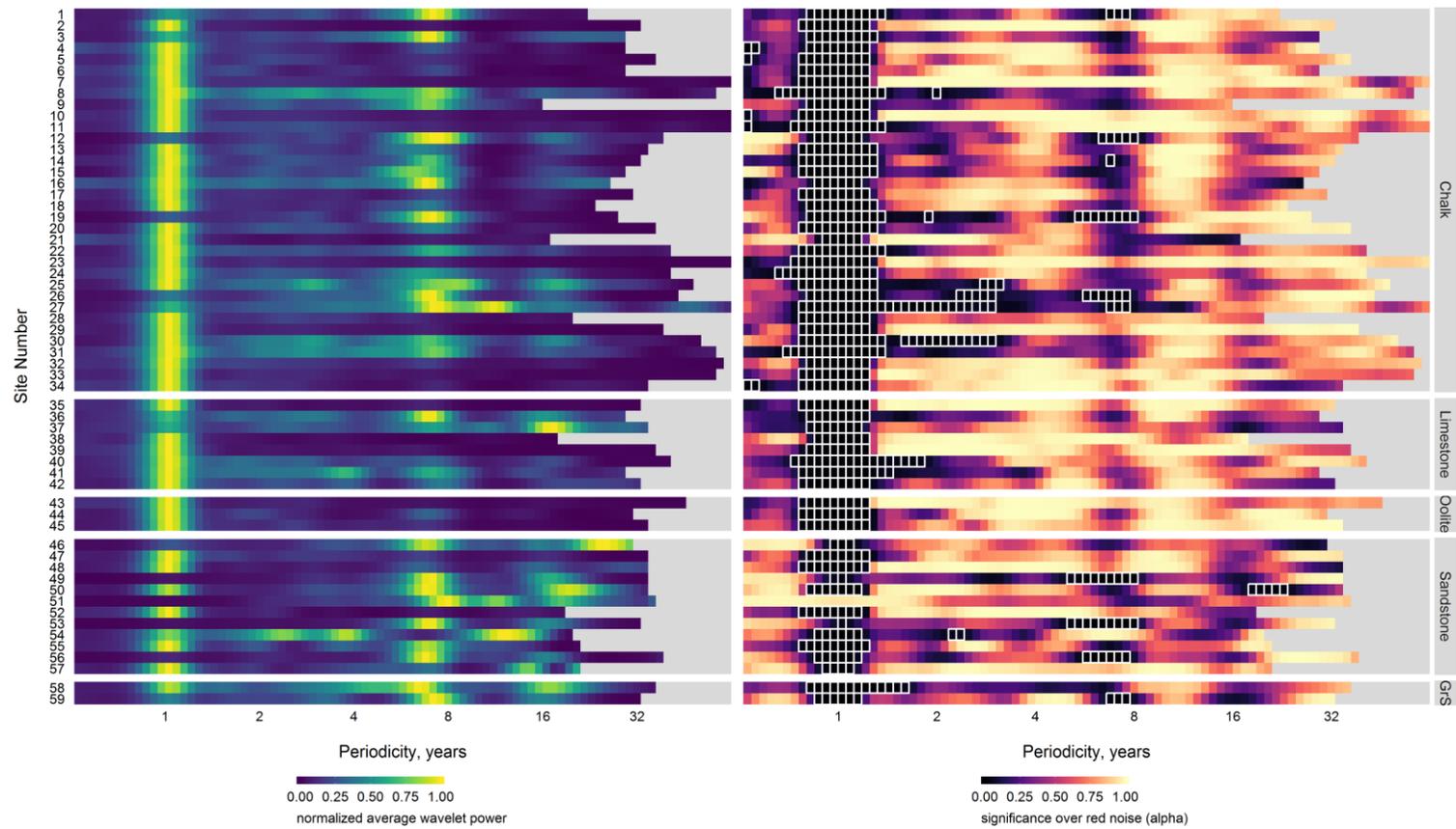
$$(x_t) = \frac{dj \cdot dt^{1/2}}{0.776 \cdot \psi(0)} \sum_s \frac{Re(W(., s))}{s^{1/2}} \quad (\text{Eq. 2})$$

209 Where dj is the frequency step and dt is the time step.

210 Negative phases of these time-reconstruction anomaly time series were compared to
211 episodes of recorded wide-scale hydrogeological drought (provided by Marsh et al. (2007) and
212 Todd et al. 2013)), to assess the relationships between ~~inter-annual~~multi-annual variability in
213 groundwater and groundwater droughts.

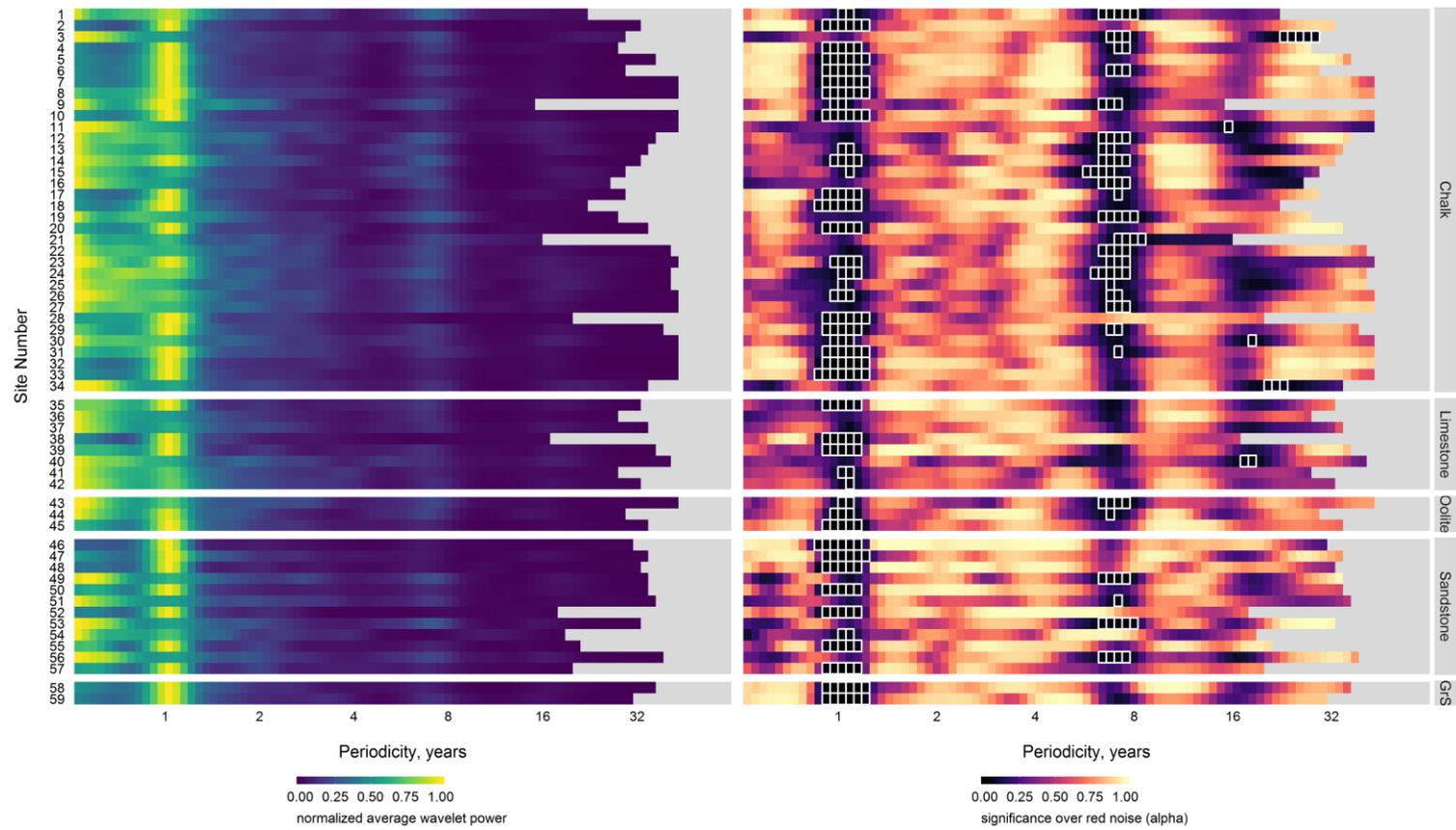
214 2.3.5 Periodicity strength quantification

215 While the wavelet power spectra from the CWT provide an estimate of the relative strength of
216 periodicities compared to the total frequency spectra, they do not provide an absolute measure
217 of a periodicities contribution to total groundwater variability (which includes noise and non-
218 periodic information). As such the percentage contributions of each time-reconstruction have
219 been calculated. Since the datasets were normalised to the standard deviation of the raw data
220 prior to the CWT, the standard deviations of the reconstructed anomaly time series represent
221 the proportion of the original standard deviation as a decimal percentage.



222

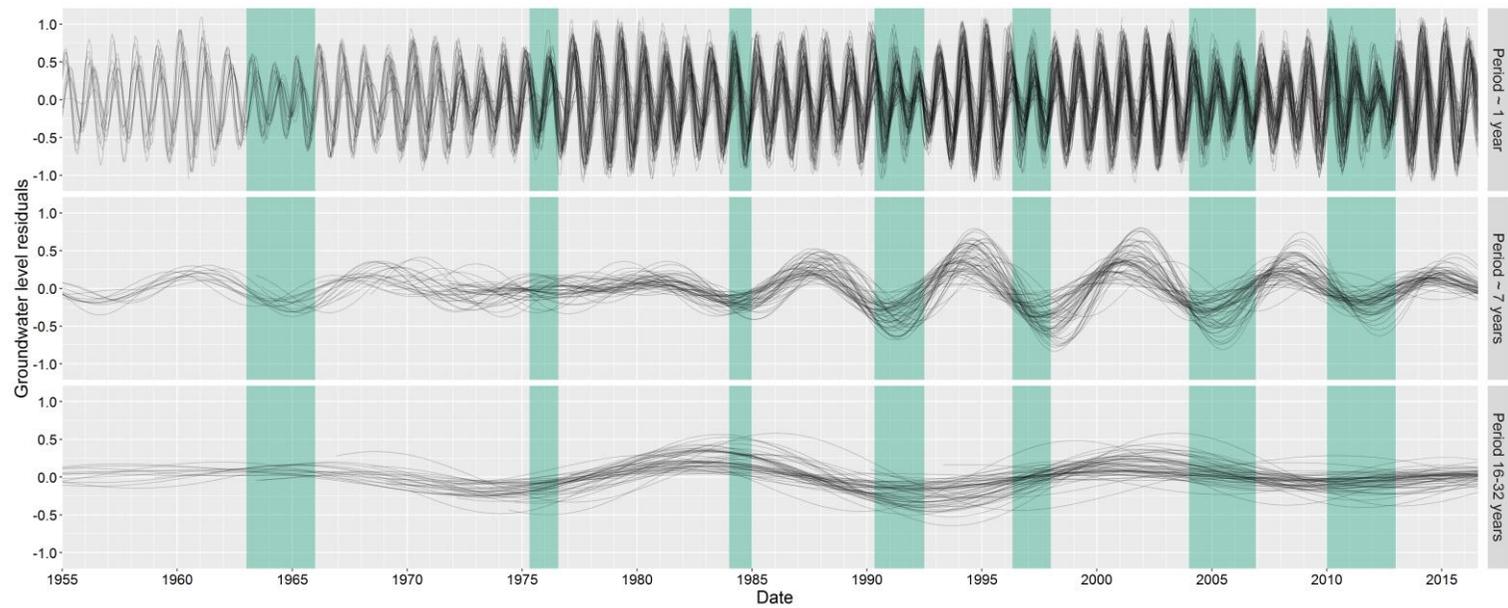
223 Figure 2 - Normalised average wavelet power spectra (left) and wavelet power significance alphas (right) for monthly groundwater levels in the 59 index boreholes (grouped by
 224 aquifer type). In the right-hand figure, boxes outlined in white are those powers that are significant over red noise to a 95% confidence interval ($\alpha \leq 0.05$).



225

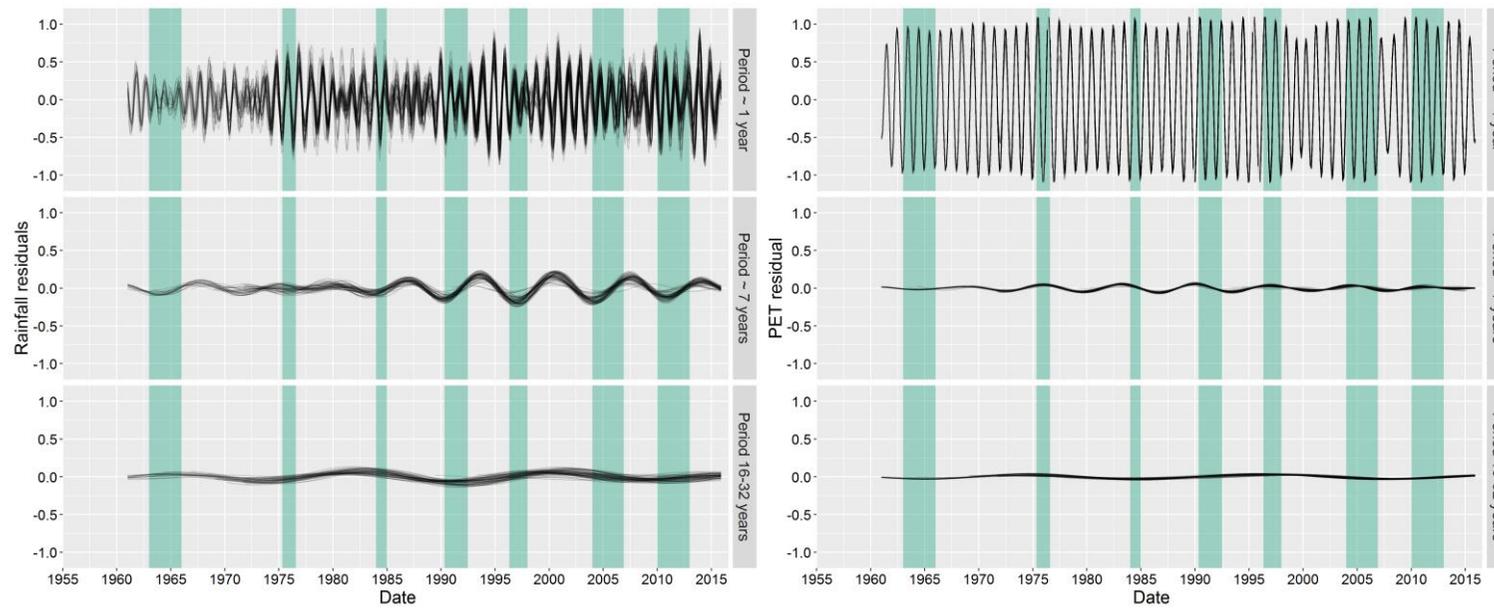
226 *Figure 3 - Normalised average wavelet power spectra (left) and wavelet power significance alphas (right) for monthly rainfall time series for co-locations of the 59 index boreholes.*

227 *In the right-hand figure, boxes outlined in white are those powers that are significant over red noise to a 95% confidence interval ($\alpha \leq 0.05$).*



228

229 *Figure 4 – Overlaid reconstructions of the three key periodic domains found across the 59 groundwater wavelet spectra are shown. All periods (both significant and non-significant)*
 230 *within these bands have been displayed to allow for comparison of period strength and phase over time. Areas shaded blue represent approximate periods of significant droughts*
 231 *in the UK. Only reconstructions between 1955 and 2017 are shown to allow clearer comparison.*



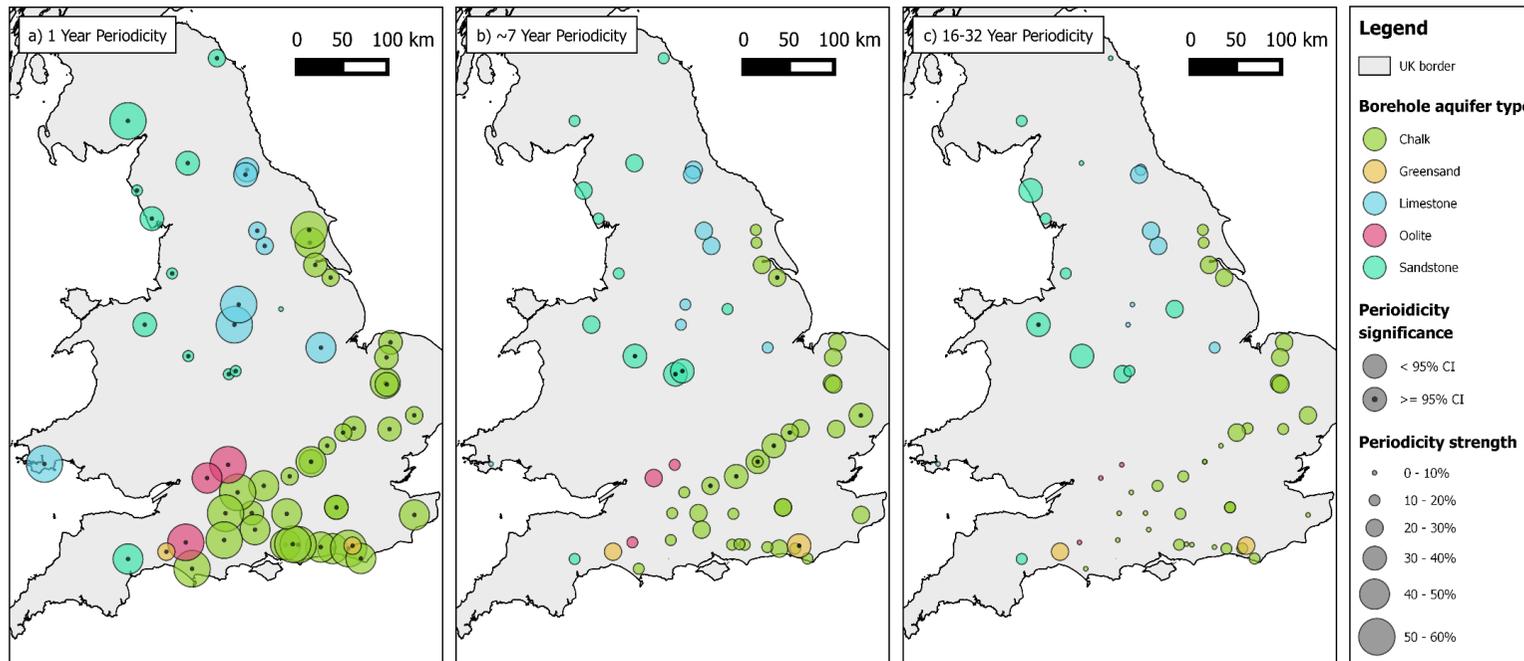
232

233 *Figure 5 – Overlaid rainfall (left) and PET (right) reconstructions of the three key periodic domains are shown. All periods (both significant and non-significant) within these bands*
 234 *have been displayed to allow for comparison of period strength and phase over time. Areas shaded blue represent approximate periods of significant droughts in the UK. Only*
 235 *reconstructions between 1955 and 2017 are shown to allow clearer comparison.*

236

237

238



239

240 *Figure 6 – Maps showing strength (percentage of the original time series standard deviation) and significance of the a) 1 year, b) ~7 year and c) 16-32 year periodicity bands.*

241 *No periodicity strength was found to be above 60% of the original signal.*

242 **3. Results**

243 **3.1. Time-averaged wavelet power and significance over red noise**

244 Wavelet power spectra (frequency strength) and alpha values (significance) for each of the 59
245 groundwater level and rainfall time series are displayed in figures 2 and 3 respectively.
246 Wavelet power is analogous to the strength of the periodicity compared to other frequencies.
247 Periodicities with alpha values less than or equal to 0.05 (95% CI) are highlighted. Bands of
248 greater wavelet power and lower alpha values at periodicities of 1, ~7 and 16-32 year(s) can
249 be seen across the majority of the groundwater and rainfall spectra for the 59 sites (herein
250 referred to as P1, P7 and P16-32 respectively). PET wavelet spectra were found to have no
251 notable or significant periodicity beyond seasonality (indicative of the UK's temperate climate),
252 and are displayed in the supplementary material.

253

254 The annual cycle (P1) exhibited the greatest power across 43 of the 59 observation borehole
255 spectra, with normalised wavelet powers ranging from 0.03 to 1 (mean of 0.84). Alpha values
256 for P1 in the observation boreholes also showed the greatest likelihood of external forcing
257 when compared to the other identified periodic domains (alpha values ranging from 0.00 to
258 0.94, mean of 0.017). All but one observation borehole (site 51) showed significant (95%)
259 alpha values for P1 wavelet power. Lower than average P1 wavelet powers were most
260 prevalent in the Sandstone lithology (6 out of 12 sandstone sites), Greensands (1 out of 2
261 sites) and to some extent, the Chalk (6 out of 35 sites). [It should be noted, however, that the](#)
262 [relatively small sample sizes of Greensand and Oolite aquifers makes interpretation of](#)
263 [systematic differences at these lithographies difficult.](#) P1 wavelet power was generally lower
264 across all the corresponding rainfall time series, which is expected given rainfall's established
265 bias towards high-frequency variability (Meinke et al. 2005). Of those boreholes with lower P1
266 power in groundwater, most (e.g. 35, 59) show greater P1 powers in rainfall (and PET)
267 indicating hydrogeological processes as the mechanism for weaker P1 periodicity. However,
268 a small number (e.g. 38, 40 and 42) had similarly low P1 periodicity in the corresponding

269 rainfall, indicating meteorological drivers for poor annual strength at these observation
270 boreholes (considering that PET showed little variance in P1 strength across the observation
271 boreholes). PET spectra and alpha values showed a universally high P1 wavelet power.

272 The second greatest wavelet power across the groundwater boreholes was between 6 and 9
273 years, roughly centred on the 7 year periodicity (P7)). Maximum normalised groundwater
274 wavelet powers ranging from 0.01 to 1 (average of 0.52) between boreholes were detected,
275 and a corresponding band of lower than average alpha values (ranging from 0.01 to 0.99,
276 mean of 0.34), indicating that this periodicity is likely to be driven by an external variance.
277 Average P7 wavelet power values were greatest for Sandstone (0.68) and Greensands (1.00),
278 and lower for Limestone (0.39) and Oolite (0.17). Chalk showed intermediate strength with the
279 greatest range (0.01 to 1.00, mean of 0.50). Ten groundwater sites showed significant (95%)
280 P7 wavelet powers (sites 1, 12, 14, 19, 26, 27, 49, 53, 55 and 59). While the P7 wavelet power
281 in the corresponding rainfall data was considerably lower than those detected in groundwater
282 level (ranging from 0.014 to 0.35, mean of 0.16), the alpha values are comparable with the P7
283 signal strength in groundwater. This indicates that P7 signals in rainfall are weak, but likely
284 driven externally. Generally lower alpha values for P7 in rainfall, compared to groundwater,
285 are also likely a result of rainfall's lower autocorrelation. Negligible wavelet powers and no
286 significance was shown at the P7 band for corresponding PET data.

287

288 The final and second mode of common ~~inter-annual~~multi-annual wavelet power was the band
289 between 16 years and 32 years (P16-32). P16-32 had an average wavelet power of 0.28
290 across all boreholes; ranging between 0.01 and 1. Similar to P7, the greatest wavelet power
291 of P16-32 was found in the Sandstone (average of 0.58) and the Greensand (average of 0.64)
292 aquifer types. Whereas Chalk, Limestone and Oolite showed relatively weaker signals
293 (averages of 0.18, 0.32 and 0.03 respectively). Only one site in the groundwater (site 50) and
294 five rainfall time series (sites 3, 11, 30, 34, 40) showed 95% significance over red noise in this
295 periodicity band.

296

297 **3.2. Reconstructed anomaly time series**

298 The three main common period domains identified by the wavelet transform (P1, ~7 and 16-
299 32 years) were reconstructed into anomaly time series using the reversed wavelet transform
300 and are presented in figure 4 for groundwater levels and figure 5 for rainfall and PET. This
301 was undertaken to allow investigation and comparison of periodic behaviour over time and to
302 assess how these reconstructed periodic signals, within multiple sites across multiple aquifers,
303 align with periods of historical groundwater drought. The behaviour of the multiple
304 reconstructed groundwater level, precipitation and PET anomaly time series (in all three
305 periodicity domains) were shown to be well-aligned in time, with positive (maxima) and
306 negative (minima) phases occurring within comparable time. The only exception to this pattern
307 was seen between 1970 and 1980 in the P7 reconstructions, where phases in the P7
308 reconstructions become misaligned. This was predominantly apparent in groundwater and to
309 a lesser extent in rainfall. Positive and negative phases of the P7 reconstructions in PET were
310 well-aligned for the entire time series.

311 Notable episodes of groundwater droughts in the UK were overlaid onto the reconstructed
312 periods in figure 5 between 1955 and 2016. With the exception of the 1975-6 event, every
313 episode of drought in this time period coincides with a negative phase of the reconstructed P7
314 groundwater anomalies. The 1975-6 drought (often used as a benchmark drought in the UK
315 due to its wide-reaching impacts (Marsh et al. 2007)) occurred at a time of notable
316 minima/maxima misalignment of the P7 period across all in-groundwater sites, and a period
317 of negative anomaly in the P16-32 reconstructions. Most recorded major droughts in the UK
318 appeared to occur irrespective of the state of the P16-32 anomaly, with droughts occurring in
319 minima and maxima of this reconstruction.

320 **3.3. Percentage standard deviation**

321 The percentage of the standard deviation in the original groundwater level signal represented
322 by each reconstructed periodicity band is shown in figure 64 for all the observation boreholes.
323 The percentages are representative of the absolute strength of the periodicity compared to
324 the recorded data variance (standard deviation).

325 P1 represents the greatest average contribution to groundwater variability across all the
326 aquifer groups (Chalk: 41%, Limestone: 40%, Oolite: 52%, Sandstone: 26%, Greensand:
327 28%). While most sites show that P1 accounts for the greatest proportion of the standard
328 deviation, P7 is the dominant periodicity at 11 of the 59 sites (5 within Sandstone, 5 within
329 Chalk and 1 within Greensand), and P16-32 is the strongest cycle in 3 of the 59 sites (3 within
330 Sandstone and 1 within Limestone). P1 strength in the Chalk appears to be greatest in the
331 South of England, with weaker strengths in the South East and East. Aside from the Chalk,
332 there are no clear spatial patterns in P1 strength. P7 accounts for an average of 21.7% of
333 signal strength across all aquifer groups, ranging from 3.8% to 40% across the observation
334 boreholes. Spatial variance in P7 signal strength is less when compared to P1, although there
335 is a noted area of significance in Chalk of South East England (e.g. the Chiltern Hills and
336 Cambridgeshire), and a smaller cluster of P7 significance in the Sandstone of the central
337 England, where the greatest P7 strengths are found. P16-32 strengths are spatially focused
338 in Eastern England for the Chalk, and the central and north-western England for the
339 Sandstone. No clear patterns for the remaining aquifer groups is apparent for the 16-32 year
340 periodicity band.

341

4. Discussion

~~The aim of this study was to assess the extent to which inter-annual cycles in groundwater levels (produced by teleconnections with climate oscillations) may be used to indicate the timing of future groundwater extremes. To achieve this, the absolute strengths of groundwater periodicities have been quantified and compared to the timing of historical droughts in the UK. In this wide-scale study, our results show for the first time that long-term cycles in groundwater levels are a crucial contributor to overall groundwater level variability. Additionally we show that much of this periodic behaviour closely aligns with episodes of historical groundwater drought over the past 60 years. These findings move beyond previous groundwater teleconnection research in the UK (Holman et al. 2011) and internationally (Kuss & Gurdak 2014; Neves et al. 2019) to provide a robust measure of the absolute contribution of inter-annual periodicities to groundwater levels fluctuations. In the following, we discuss the findings presented in this paper within the context of the research objectives and the implications for improved water resource management.~~

4.1. Characterisation of signal presence and strength in groundwater level

Many studies have focused on the role of seasonality in defining groundwater variability, and the onset and severity of groundwater drought (Jasechko et al. 2014; Hund et al. 2018; Mackay et al. 2015; Ferguson & Maxwell 2010). While we show that the annual cycle is an important component of groundwater response, it is often not representative of overall behaviour, accounting for (on average) less than half of total groundwater level variability. Conversely, we show that ~~inter-annual~~multi-annual periodicities form an unprecedented proportion of total groundwater variability; with 41% of sites (24 out of 59) exhibiting ~~inter-annual~~multi-annual periodicity strength that is comparable to (within 10%), or greater than, seasonality. It is expected that the strength of ~~inter-annual~~multi-annual cycles in groundwater level will vary according to signal strength in recharge drivers (e.g. rainfall and evapotranspiration) and hydrogeological processes that lag or attenuate long-term changes in these recharge signals (Van Loon 2013; Van Loon 2015; Townley 1995; Dickinson et al.

2014). These two processes may explain the local differences in signal strength between sites in aquifer types and geographically across the UK, as displayed in our results. For instance, pronounced ~~inter-annual~~multi-annual variability (significant 7 year cycles and stronger 16-32 year cycles) in the Chalk sites is generally associated with catchments of thicker unsaturated zones, larger interfluves or areas of weaker corresponding seasonality in rainfall (for example, the Chiltern Hills in South East England). These catchment properties have been shown to dampen higher frequency variability between rainfall and groundwater response due to storage buffers, thereby producing a sensitivity to ~~inter-annual~~multi-annual variability (Peters et al. 2006; Van Loon 2013). ~~Inter-annual~~Multi-annual cycles are also generally strong for the granular porosity aquifers (Sandstone and Greensand); which is to be expected given the influence of lower hydraulic diffusivity (typical of granular porosity flow) on the suppression of high-frequency variability (Townley 1995). This also agrees with Bloomfield & Marchant (2013) who document sensitivity to long-term accumulation in rainfall in UK Sandstone aquifers. Conversely, the Limestone and Oolite aquifer types exhibit weaker ~~inter-annual~~multi-annual periodicities in groundwater level, with strong seasonality. Townley (1995) and Price et al. (2005) document that, due to their faster-responding fracture porosity with low storativity, limestone lithographies have a lower damping capacity of finer-scale variability in recharge, meaning they are able to respond in-time to the strong seasonality in PET and rainfall. ~~We demonstrate that inter-annual periodicity strength in groundwater level is the result of both meteorological (principally rainfall) and hydrogeological processes. Our f~~results typically show lower percentage contributions of multi-annual periodicities to total groundwater level variability than some previous ~~internatona~~international studies (Kuss & Gurdak 2014; Neves et al. 2019; 389; Velasco et al. 2015). This can be explained by potentially weaker periodicities in driving climatic circulations over Europe compared to North America (as indicated by Kuss & Gurdak (2014)), or UK-specific hydrogeological properties such as smaller aquifer size compared to North America or continental Europe, which may affect teleconnection strength (Rust et al, 2018). However, we also expect lower percentage contributions of multi-annual periodicities due to our use of unmodified groundwater level datasets prior to spectral

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decomposition (wavelet transform). Using unmodified level data has enabled us to represent the absolute contribution of multi-annual variability to groundwater level behaviour at each site.

4.2. Evidence for teleconnection control on ~~inter-annual~~multi-annual groundwater variability

Here, we discuss the evidence that the ~~inter-annual~~multi-annual variability present in UK groundwater level records (as previously discussed) is the result of teleconnection influences with climatic oscillations. The conceptualisation of groundwater teleconnections of Rust et al (2018) suggests that a teleconnection between the oscillatory climate systems and groundwater level would be associated with;

- a) an apparent and coherent ~~inter-annual~~multi-annual periodicity band within groundwater sites across a wide geographical area, that aligns with known ~~inter-annual~~multi-annual variability in indices of climatic oscillations (for instance, the 7-year periodicity of the NAO (Hurrell et al. 2003),
- b) increased likelihood that this periodicity band is the result of an external influence, and not the result of internal red-noise variability of the groundwater level time series (as indicated by Allen et al. (1996) and Meinke et al. (2005))
- c) comparable signals in rainfall as established drivers for ~~inter-annual~~multi-annual groundwater variability, and
- d) broad alignment of minima and maxima of time-reconstructed ~~inter-annual~~multi-annual periodicities. Some fine-scale misalignment in groundwater periodicities is expected as a result of unsaturated and saturated zone lags between rainfall and groundwater response (Van Loon 2013; Peters et al. 2006; Dickinson et al. 2014; Cuthbert et al. 2019).

The majority of groundwater level hydrographs and corresponding rainfall profiles showed a coherent band of increased periodicity strength and periodicity significance principally around the 7-year frequency range and, to a lesser extent, the 16-32 year range. The 7 year periodicity closely compares to the principle 7-year periodicity documented in the strength of the NAO's atmospheric dipole, which has been associated with ~~inter-annual~~multi-annual periodicities in rainfall (Meinke et al. 2005) and groundwater globally (Tremblay et al. 2011; Kuss & Gurdak 2014; Holman et al. 2011; Neves et al. 2019). Additionally, the time-reconstructions show clear temporal alignment of minima (with the exception of the 1975-6 period, which will be discussed later), indicating the wide-spread coherent influence of a climatic teleconnection. As such, we corroborate with existing research that documents the control of the NAO on UK rainfall (Alexander et al. 2005; Trigo et al. 2004), and show new evidence of the wide-spread propagation of ~~inter-annual~~multi-annual variability in rainfall through to spatio-temporal ~~inter-annual~~multi-annual groundwater variability, conceptualised by Rust et al (2018).

While the NAO is known to be the dominant mode of winter climate variability in Europe (López-Moreno et al. 2011; Alexander et al. 2005; Hurrell & Deser 2010), the second strongest is provided by the East Atlantic (EA) pattern (Wallace & Gutzler 1981). The EA is similar in frequency structure to the NAO but shifted southward, however it has been shown to exhibit its own internal variability (Hauser et al. 2015; Tošić et al. 2016; (Moore et al., 2013). Importantly, the EA has been shown to exhibit a 16-32 year periodicity (Holman et al, 2011), and therefore aligns with the second strongest mode of ~~inter-annual~~multi-annual variability in groundwater and rainfall documented in this study. As such, the increased strength, significance and minima-alignment of the 16-32 year periodicity range detected in groundwater levels in this paper may be explained through a teleconnection between the EA pattern and European winter climate variability. Similar to the 7-year periodicity, the 16-32 year cycle detected in groundwater levels shows an increased likelihood of external variance, and temporal alignment of minima and maxima when reconstructed back into the time domain. As such, we consider the EA to be the ultimate driver of the 16-32 year periodicity detected in UK

~~groundwater level~~. While the EA has received little focus in climate variability research compared to the NAO, our findings here ~~align-supportwith~~ Krichak & Alpert (2005) who document a multi-decadal control on UK and European precipitation through shifting phases of the EA, and Holman et al (2011) who detected weak relationships between the EA and groundwater levels in the UK. Comas-Brua and McDermotta (2014) suggest that much of the multi-decadal climate variability (temperature and precipitation) in the North Atlantic region can be explained by a modulation of the NAO by the EA, which may contribute to the spatial and temporal variability seen in both the ~7 year and 16-32 year reconstructions across the borehole sites. In summary, the modes of multi-annual variability detected in the majority of UK groundwater level hydrographs and rainfall time series appear to be best explained via a teleconnection with the NAO and EA's principle periodicities.

~~we assert that the inter-annual variability detected in UK groundwater and rainfall data is likely the result of a climatic teleconnection with both the NAO and the EA. As such, we document the first evidence of the absolute strength of both the NAO and the EA's control on 7-year and 16-32 year variability in groundwater systems respectively.~~

4.3. Teleconnections as indicators for groundwater extremes

The final objective of this paper was to assess the extent to which the timing of ~~inter-annual~~multi-annual periodic groundwater level oscillations align with the timing of recorded groundwater droughts. To achieve this, documented periods of groundwater drought have been compared to reconstructed periodicities within groundwater level. We show principally, that every documented groundwater drought between 1955 and 2014 ~~aligns with~~occurs during a negative phase of the ~7 year cycle detected in the majority of UK groundwater boreholes, with the exception of the 1975-6 drought. ~~In addition to the strength of a ~7 year cycle in UK groundwater level previously discussed, this alignment provides strong evidence that the NAO influences inter-annual groundwater variability, resulting in groundwater drought on an approximate 7-year recurrence. As mentioned, the only drought that does not fit this~~

~~pattern is the 1975-6 drought, which occurred during the only episode of temporal misalignment in the reconstructed 7-year groundwater level periodicities.~~

Groundwater droughts are typically the result of multi-annual accumulation of rainfall deficits, with the specific timing and duration of drought (for a particular site) also driven by sub-annual rainfall and evapotranspiration (Van Loon et al. 2014; Peters 2003). Marsh et al. (2007) identify a multi-annual decline in rainfall for the majority of droughts in the UK over the past 60 years, with rainfall deficits reaching a critical accumulation period of a 2-3 years in the lead-up to drought commencement (Folland et al. 2015). It is therefore to be expected that the majority of droughts are captured within the negative phases of the ~7 year cycle in the groundwater level anomalies, as this cycle (along with the 16-32 year cycle) represents groundwater's multi-annual response to the land-atmosphere water flux. The 1975-6 drought does not fit this pattern as we show a clear disruption to the ~7 year cycle during this period. This is of particular interest as this event is generally acknowledged to be anomalous for the UK. The severity of this event has been solely attributed to a short-term meteorological state (i.e. high-pressure atmospheric blocking) in existing literature, with very little long-term decline in groundwater levels (Rodda & Marsh, 2011; Bloomfield & Marchant 2013b). It is therefore therefore to be expected that the 1975-6 drought did not occur during a coherent negative phase of the ~7 year cycle detected in groundwater levels, and that we see a more pronounced suppression of seasonality during this time. Furthermore, we note that most droughts do not perfectly align with the minima of this ~7 year cycle. As previously stated, the commencement of a drought is dependent a combination of on multi-annual land-atmosphere water fluxes and particular sub-annual hydrological conditions (Van Loon et al. 2014). Therefore, this ~7 year fluctuation could be considered a cycle of increased drought risk, where groundwater resources may be more sensitive to sub-annual hydrological conditions. The 16-32 year periodicity, while also a representation of groundwater's multi-annual response to moisture balance, represents a smaller proportion of total groundwater behaviour and as such appears

less representative of drought timings. Despite this, it is likely that this signal still has a role in modulating the severity of the ~7 year component.

The NAO and EA's control on long-term rainfall deficits in the UK and Europe has already previously been identified by many studies (López-Moreno et al. 2011; Fowler & Kilsby 2002; Hurrell 1995). Here, we provide evidence to suggest that the NAO teleconnection with long-term rainfall volumes, in particular, propagates to detectable modes of groundwater level behaviour, creating episodes of increased drought risk in-line with the NAO's principle periodicity of approximately 7 years. While the effects of (non-) stationarity between the NAO, EA and UK hydrogeology have not been assessed in this study, these detected cycles may yield improved foresight into future episodes of increased drought risk in the UK. This is especially important given the proportion of groundwater level variability these cycles represent. We also note that teleconnections are not persistent and can be disrupted as exemplified by the 1975-6 drought. Our findings here agree with Parry et al (2011) who found no relationship with this drought and the NAO phase or strength. Peings & Magnusdottir (2014) suggest that atmospheric blocking prohibits the expected effects of the NAO on UK and European rainfall, which may explain both the 1975-6 drought and the disruption to the 7-year periodicity in UK groundwater (Rodda & Marsh 2011). This therefore further highlights the importance of atmospheric blocking in regulating groundwater variability in the UK (Shabbar et al. 2001). The 1975-6 drought is of particular interest for the UK as it is often used as a benchmark drought, being one of the most severe droughts in recent history (March et al. 2007). (Rodda & Marsh 2011) attributed the severity of this drought to several short-term influences (such as positive pressure anomalies driving dryer conditions), in contrast to the multi-year accumulation of moisture deficits that typically result in hydrogeological drought, particularly in the UK (Van Loon 2015; Bloomfield & Marchant 2013). Our results periodic time reconstructions reiterating this, showing a disruption to the, otherwise, strong 7-year fluctuation in groundwater level during 1975-6, in addition to an intense short-term suppression

of seasonality. As such, these results infer that the NAO did not directly modulate the 1975-6 drought, agreeing with Parry et al (2011) who found no relationship with this drought and the NAO. This potentially points to both the 1975-6 drought and the disrupted NAO being modulated in parallel by a wider atmospheric control. Peings & Magnúsdóttir (2014) suggest that atmospheric blocking prohibits the expected effects of the NAO on UK and European hydrology, which may indirectly explain both the 1975-6 drought and the disruption to the 7 year periodicity in UK groundwater (Rodda & Marsh 2011).

While the 16-32 year periodicity in groundwater level does, in general, align with historical recorded droughts, this is not as coherent as with the 7 year cycle, with droughts occurring in positive and negative phases. However, given the percentage contributions of the 16-32 year periodicity to total groundwater variability, it is likely that this signal has a role in modulating the severity of droughts influenced by the NAO 7-year drought cycle (as suggested by Comas-Brua and McDermotta (2014)). For instance, the severe 1975-6 drought occurred at a slight negative phase of the 16-32 year cycle, indicating that a portion of the severity of this drought was the result of EA's influence.

Based on the alignment of the 7 year cycle (and partial alignment of the 16-32 year cycle) with historical recorded UK droughts, we conclude that the NAO (and EA) directly modulate the severity and timing of droughts in the UK. Furthermore, the 7-year cycle is shown to be a sufficient indicator of the onset of droughts, based on this historical alignment. Consequently, this 7-year cycle can be extrapolated beyond the end of the dataset used in this study. Therefore, based on a projected 7-year cycle, we predict the UK will likely enter drought conditions around 2018/19, 2025/6 and 2033/4, assuming the continuation of the NAO system's influence. This projection is further validated by the onset of drought conditions in the UK in mid-2018 (Hannaford, 2018).

~~In the UK, the economic regulator has implemented several measures to promote the trading of water between water supply companies to enable a more robust water supply system (Ofwat 2010; Deloitte LLP 2015). Here, we show that recursive patterns in groundwater contribute to a considerable proportion of the total groundwater level variability and therefore may provide new insights to allow undertakers of water supply to trade water further into the future, depending on teleconnection sensitivities. Such forecasted planning could help to reduce the ecological and human impacts of groundwater drought by allowing more time to plan and organise the required water transfers from areas less susceptible to teleconnection-driven drought.~~

5. Conclusions

This paper assesses the role of ~~inter-annual/multi-annual~~ variability and ocean-atmosphere systems in influencing groundwater drought. We quantify, ~~for the first time globally,~~ the absolute contribution of ~~inter-annual/multi-annual~~ cycles to groundwater variability, and provide new evidence for the influence of the NAO's control of European rainfall on UK groundwater drought over the past 60 years.

The wavelet transformation was used to identify and evaluate bands of periodic external influence on UK groundwater level hydrographs. ~~We,~~ documenting the ~~strength of a 1-, approximate 7 and 16-32 year cycle in the majority of sites assessed~~ strength of multi-annual behaviour that align with the NAO's principal periodicity (approximately 7 years) and the EA's principal periodicity (16-32 years). We find that seasonality accounts for an average of 39% of groundwater level variance across boreholes; with 7-year cycle accounting for an average of 21%, and 16-32 years accounting for 15%. Furthermore, we show the majority of UK droughts align with a negative phases of the 7-year cycle indicating periods of increase drought risk as part of this periodicity. ~~Furthermore, the minima of NAO-driven cycles in groundwater level~~

~~align with the occurrence of recorded groundwater drought, allowing the estimation of future drought occurrences on a multi-annual timescale. The analysis demonstrates that the NAO is the principle control (and the EA as the secondary control) on inter-annual variability in UK groundwater level, and provides a new approach to forecast the onset of groundwater droughts through an extrapolation of cyclical behaviour into the future. As such we identify 2018/19, 2025/6 and 2033/4 as likely episodes of future droughts in the UK. Although further work is required to better understand the teleconnection sensitivity, the methods described in this paper provide a robust and transferable approach for assessing the quantitative influence of teleconnections in hydrological datasets. It is clear from our results that long-range groundwater drought forecasts via climate teleconnections present transformational opportunities to drought prediction and its management across the North Atlantic region.~~

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The groundwater level data used in the study are from the WellMaster Database in the National Groundwater Level Archive of the British Geological Survey. The data are available under license from the British Geological Survey at <https://www.bgs.ac.uk/products/hydrogeology/WellMaster.html> (last access: 26/03/2019).

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