



1	A 424-year tree-ring based PDSI reconstruction of <i>Cedrus deodara</i> D. Don from Chitral
2	HinduKush Range of Pakistan: linkages to the ocean oscillations
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19	Abstract. Currently, the rate of global warming has led to persistent drought patterns. It is
20	considered to be the preliminary reason affecting socio-economic development under the
21	background of dynamic forecasting of water supply and forest ecosystems in West Asia.
22	However, long-term climate records in the semi-arid Chitral mountains of northern Pakistan are

23 seriously lacking. Therefore, we developed a new tree-ring width chronology of *Cedrus deodara*



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24	spanning the period of 1537-2017. We reconstructed the March-August Palmer Drought
25	Sensitivity Index (PDSI) for the past 424 years back to A.D. 1593. Our reconstruction was
26	featured with nine dry and eight wet periods 1593-1598, 1602-1608, 1631-1645, 1647-1660,
27	1756-1765, 1785-1800, 1870-1878, 1917-1923, 1981-1995, and 1663-1675, 1687-1708, 1771-
28	1773, 1806-1814, 1844-1852, 1932-1935, 1965-1969 and 1996-2003, respectively. This
29	reconstruction is consistent with other dendroclimatic reconstructions in west Asia, confirming
30	its reliability. The analysis of the Multi-Taper Method and wavelet analysis revealed drought
31	variability at periodicities of 2.1-2.4, 3.3, 6, 16.8, and 34-38 years. The drought patterns could be
32	linked to the broad-scale atmospheric-oceanic variability such as El Niño-Southern Oscillation
33	(ENSO), Atlantic Multi-decadal Oscillation (AMO) and solar activity. In terms of current
34	climate conditions, our findings have important implications for developing drought-resistant
35	policies in communities on the fringes of Hindu Kush mountain Ranges in northern Pakistan.
36	Keywords: Tree rings, Growth-climate response, Drought variability, ENSO,
37	dendroclimatology, the broad-scale atmospheric-oceanic variability
38	
39	1. Introduction
40	Numerous studies have shown that the intensity and frequency of drought events have
41	increased due to rapid climate warming (Stocker et al., 2013; IPCC, 2013). The apparent drought
42	has had serious adverse effects on social, natural, and economic systems (Ficklin et al., 2015; Yu
43	et al., 2018; Yao et al., 2015). Global drought is considered to be the most destructive climate
44	disaster that has caused billions of dollars in worldwide loss (Zhu et al., 2009; Seneviratne et al.,

46 unsustainable water supply (Hellmann et al., 2016; Wang et al., 2017). It is not only critical for

2

2012). The accelerated rate of warming changes the regional hydrological regimes, leading to an





47	agricultural production, but also leads to forest mortality, vegetation loss (Martínez-Vilalta and
48	Lloret, 2016), increases risk of wildfires (Turner et al. 2015; Abatzoglou and Williams 2016),
49	and affects social stability and well-being of the people. The degradation of grassland and loss of
50	livestock caused by drought will eventually affect the lifestyle of nomadic peoples, especially in
51	high-altitude forested areas (Shi et al., 2019; Pepin et al., 2015; Wu et al., 2015). In high altitude,
52	arid and semi-arid areas, forest growth is sensitive to climate change, so there is a need to
53	understand more about the long-term drought regimes of the past.
54	Northern Pakistan belongs to the Indus civilization and occupies an important position in
55	the history of drought-related culture. In addition, Pakistan has a semi-arid climate, and its
56	agriculture economy is most vulnerable to drought (Kazmi et al., 2015; Miyan, 2015). The
57	prolonged drought from 1998 to 2002 reduced the production from 80% to 60% (Ahmad et al.,
58	2004). Trees in the Chitral, Hindu-Kush Himalayan (HKH), northern Pakistan are important for
59	climate reconstruction because of their distinct and complex topography, high altitudes,
60	inadequate availability of metrological record, shadow precipitation (out of summer monsoon
61	reach), and unique precipitation seasonality (Khan et al., 2010). Meanwhile, a large hydroelectric
62	reservoir in Pakistan, such as Tarbela Dam, is built on the Upper Indus River (UIB), which
63	receives water in the form of melting glaciers from HKH (Rao et al., 2018; Rashid et al., 2018).
64	In addition, the Chitral forest, an essential archive of dendroclimatic research, has been heavily
65	felled. According to the Pakistani Government, only 4% of the country is still forested, compared
66	with 2% reported by the Food and Agriculture Organization. In northern Pakistan, there is little
67	research on the dendroclimatology (Treydte et al. 2006), and instrumental climate records are
68	inadequate in terms of quality and longevity. Climate reconstruction is urgently needed in
69	northern Pakistan in order to understand long-term climate change, expand climate record,





70	develop forest management strategies, and understand the impact of past climate change.
71	Researchers used multiple proxies, including ice cores, speleothems, lake sediments, and
72	historical documents and tree rings to reconstruct past short-term or long-term climate change.
73	Among them, tree growth is susceptible to evapotranspiration, soil and air temperature, air
74	humidity, and soil moisture (Paudel and Vetaas, 2014). Therefore, the study of tree rings is
75	highly recommended for dendroclimatic studies.
76	Meanwhile, tree rings were widely used in long-term paleoclimatic reconstructions and
77	forecasting future climate because of their accurate dating, high resolution, wide distribution,
78	easy access, long time series, and abundant environmental information recorded (Zhang et al.,
79	2016; Chen et al., 2019; Shi et al., 2018; Klippel et al., 2017; Esper et al., 2016)
80	The intensity and frequency of the drought trend are very complex in HKH, and there is
81	no manifest measuring tool to compute how long the drought period might persist. Climate
82	uncertainty complicates the situation, for example, whether the drought trend is increasing or
83	decreasing (Chen et al. 2019). Most studies believe that the wetting trend in HKH is going to
84	increase in current decades (Treydte et al. 2006). However, some extreme drought events in the
85	region are very severe and persistent (Gaire et al., 2017). Little work had been done to examine
86	the linkages of the drought trend and driving forces (Gaire et al., 2017; Cook et al., 2003). The
87	Palmer drought sensitivity index (PDSI) is a preferred index for reconstructing past climate
88	records (Ficklin et al. 2015; Schrier et al 2013). PDSI has been widely used in the world,
89	including Australia (O'Donnell et al., 2018), New Zealand (Palmer et al., 2015), Africa, Europe,
90	and Asia (Cook et al., 2015; Ficklin et al., 2015, Tejedor et al., 2017; Steinschneider et al.,
91	2018). In the United States, federal and state agencies also use PDSI for water resources
92	management (Werick et al., 1994)





93	In this study, we collected drought-sensitive tree-ring cores of Cedrus deodar from the
94	upper and lower Chitral, Hindu-Kush Himalayan (HKH) region of Pakistan. These tree rings
95	have a good potential for dendroclimatic study (Yadav, 2013). Then, March-August PDSI was
96	reconstructed for the past 400 years to examine the climatic variability and driving forces. To
97	verify its reliability, we compared our reconstructed PDSI with other available paleoclimatic
98	records near our research area. The intensity and drought mechanism in this area were also
99	discussed. This will be the first time that the drought index has been reconstructed in northern
100	Pakistan and is considered to be a baseline for more tree-ring reconstruction in Pakistan.
101	
102	2 Material and Methods
103	2.1 Study area
104	We conducted our research in the Chitral, Hindu-Kush (HK) Mountains of northern
105	Pakistan (35.36° N, 71.48° E, Fig. 1). Northern Pakistan is a subtropical monsoon climate.
106	Summer is dry and hot, spring is wet and warm, and in some high altitudes it snows all the year
107	round. March is the wettest month (with an average precipitation of 107 mm), while July or
108	August is the driest (with an average precipitation of 6.3 mm). July is the hottest month (mean
109	monthly temperature 36 °C), and January is the coldest (mean monthly temperature -0.8 °C) (Fig.
110	2). The soil at our sampling sites is acidic, with little variation among stand of forest. Similarly,
111	the soil water holding capacity ranged from $47\% \pm 2.4\%$ to $62\% \pm 4.6\%$, while the soil moisture
112	ranged from 28%±0.57% to 57%±0.49% (Khan et al., 2010). Chitral forest is mainly composed
113	of Cedrus deodara, Juglans regia, Juniperus excelsa, Quercus incana, Q. diltata, Q. baloot and
114	Pinus wallichiana. C. deodara was selected for sampling because of the high dendroclimatic
115	value (Khan et al., 2013).





116	Climate data such as monthly precipitation and temperature obtained from the
117	meteorological station of Chitral, northern Pakistan during the period of 1965-2013. PDSI
118	downloaded from the datasets of nearest grid point (35.36° N, 71.48° E) through Climatic
119	Research Unit (CRU TS.3.22, 0.5° latitude $\times 0.5^{\circ}$ longitude). The most common reliable period
120	spanning from 1960-2016 was used (http://climexp.knmi.nl/) for dendroclimatic studies (Harris
121	et al., 2014; Yadav et al., 2014; Shekhar, 2014).
122	
123	2.2 Tree rings collection and chronology development
124	Tree-ring cores collected in Chitral forest from Cedrus deodar trees, located in northern
125	area of Pakistan. To maintain the maximum climatic signals contained in tree rings, undisturbed
126	open canopy trees were selected. One core per tree at breast height (~1.3 m above ground) was
127	sampled using a 5.15 mm diameter increment borer (Haglöf Sweden, Langsele, Sweden). In
128	addition, several ring width series were also downloaded from the International Tree-Ring Data
129	Bank from Bumburet forest and Zairat forest (https://www.ncdc.noaa.gov/paleo-search/)
130	collected in 2006 (Fig. 1).
131	All tree-ring samples were first glued than progressively mounted, dried, and polished
132	according to a set procedure (Fritts, 1976; Cook and Kairiukstis, 1990). The preceding calendar
133	year was assigned and properly cross-dated. False rings were identified by Skeleton-plot and
134	crossdated as mentioned in Stokes and Smiley (1968).
135	The cores were measured using the semi-automatic Velmex measuring system (Velmex,
136	Inc., Bloomfield, NY, USA) with an accuracy of 0.001 mm. The COFECHA program was then
137	used to check the accuracy of cross-dating and measurement (Holmes, 1983). All false one has
138	been modified and the cores that didn't match the master chronology were not used to develop





139	tree-ring chronology. For quality checks, the COFECHA 2002 program was used. (Holmes,
140	1998). The synthesized tree-ring width chronology (Fig. 3) was built by the program R (Zang
141	and Franco, 2015). To preserve climate signals and avoid noise, appropriate detrending was
142	introduced. Biological trends of tree growth associated with tree age were conservatively
143	detrended by fitting negative exponential curves or linear lines (Fritts, 1976). The tree-ring
144	chronology was truncated where the expressed population signal (EPS) was large than 0.85,
145	which is a generally accepted standard for more reliable and potential climate signal results
146	(Wigley et al., 1984; Cook and Kairiukstis, 1990). The mean correlation between trees (Rbar),
147	means sensitivity (MS) and EPS were calculated to evaluate the quality of chronology (Fritts,
148	1976). Higher mean sensitivity and EPS were considered to be strong response to climate change
149	(Cook and Kairiukstis, 1990).
150	2.3 Statistical analysis
151	Correlation analysis was conducted between tree-ring indexes (TRI) and monthly
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162	incoherence, which is often used in previous studies (Fritts, 1976; Cook et al., 2010). DWT is
163	used to calculate first-order autocorrelation or linear trend in regression residuals (Cook and
164	Pederson, 2011; Wiles et al., 2014). The RE and CE larger than zero were considered skill
165	(Fritts, 1976; Cook and Kairiukstis, 1990).
166	We defined the dry or wet years of our reconstruction with the PDSI value greater than or
167	less than the mean ± 1 standard deviation. We assessed the dry and wet periods for many years
168	based on strength and intensity.
169	Although there were a few reconstructions in our study area, we still compared our
170	reconstruction with other available drought reconstructions near the study area (Treydte et al.,
171	2006). The Multi-taper Method (MTM) was used for spectral analysis, and the Wavelet analysis
172	was used to determine the statistical significance of band-limited signals embedded in red noise
173	by providing very high-resolution spectral estimates that eventually give best possible option
174	against leakage. To identify the local climate change cycle, background spectrum was used
175	(Mann and Lees, 1996).
176	
177	3 Results
178	3.1 Main climate limiting factors for Cedrus deodar

179 The statistical parameters of the tree-ring chronologies, including MS (0.16), Rbar (0.59),

and EPS (0.94) indicated that there were enough common signals in our sampled cores, and our

181 chronology is suitable for dendroclimatic study. According to the threshold of EPS (5 trees or

182 EPS > 0.85), the 1593-2016 was selected as the reconstruction period to truncate the period

183 1537-1593 of the chronology (Fig. 3).





184	The TRI was significantly positively correlated with monthly PDSI ($p < 0.01$) (Fig. 4a).
185	While it was positively correlated with precipitation in the previous October and the current
186	February-May and significantly negatively correlated ($p < 0.001$) with precipitation in the
187	previous September. TRI was significantly positively ($p < 0.001$) with the minimum temperature
188	in the previous September, December, and the current year January and February (Fig.4b).
189	Similarly, TRI was significantly negatively ($p < 0.001$) correlated with the mean maximum
190	temperature in the previous January, October, December and the current February, March, May,
191	and June, and significantly positively correlated with the mean maximum temperature in
192	September.
193	3.2 Reconstruction of the past drought variation in northern Pakistan
194	Based on the above correlation analysis results, the March-August PDSI was the most
195	suitable for seasonal prediction and reconstruction. The linear regression model between TRI and
196	mean March-August PDSI for the calibration period from 1960 to 2016 was significant (F =52.3,
197	$p < 0.001$, adjusted $R^2 = 0.47$, $r = 0.69$). The regression model was:
198	Y = 5.1879x - 5.676
199	Where Y is the mean March-August PDSI and x is the tree-ring index.
200	The split calibration-verification test showed that the variance explained was higher during
201	the two calibration periods (1960-1988 and 1989-2016). Statistics of R , R^2 , ST, and PMT are all
202	significant at $p < 0.05$, indicating that the model was reliable (Table 1). In addition, the most
203	rigorous RE and CE tests in the verification period were all positive. For the calibration period
204	(1960-2016), the reconstruction accounted for 39.2% of the scPDSI variation (37.6 after
205	accounting for the loss of degrees of freedom). These results made the model more obvious and
206	robust in scPDSI reconstruction.
	9





207	The instrumental and reconstructed scPDSIs of the Hindu Kush Mountains had similar
208	trends and parallel calibrations during short- and long-time scales in the 20th century (Fig 5).
209	However, the reconstructed scPDSI did not fully capture the magnitude of extremely dry or wet
210	conditions.
211	
212	3.2 The drought regime in the Chitral Mountain, northern Pakistan for the past 424 years
213	Fig. 5 showed the dry and wet years of the past 424 years (1593-2016) in the Chitral
214	Mountain in northern Pakistan. The dry periods were recorded in 1593-1598, 1602-1608, 1631-
215	1645, 1647-1660, 1756-1765, 1785-1800, 1870-1878, 1917-1923, and 1981-1995. Similarly, the
216	wet periods were recorded in 1663-1675, 1687-1708, 1771-1173, 1806-1814, 1844-1852, 1932-
217	1935, 1965-1969, and 1996-2003. Spectral analysis showed that the historical PDSI changes in
218	the Hindukush Mountains showing several significant (95% or 99% confidence level) with
219	periods at 33-38 (99%), 16.8 (99%), 2-3 (99%) years, corresponding to significant periodic peaks
220	(Fig. 7)
221	
222	4 Discussion
223	4.1 Drought variation in Chitral HinduKush Range of Pakistan
224	The growth-climate relationship revealed the positive and negative influence of
225	precipitation and summer temperature on growth. It means that water availability (PDSI) is the
226	main limiting factor affecting the growth of C. deodara. Singh et al. (2006) reported that the
227	previous October precipitation limited the growth of C. deodara, while Ahmed et al. (2011)
228	found no such effect. Except for last August, November, and current September, maximum
229	temperature has a negative impact on the growth of C. deodara, while the minimum temperature





230	does not. These results suggest that moisture conditions in April-July are critical to the growth of
231	C. deodar in the study area (Hussain et al., 2007; Khan et al., 2013; Borgaonkar et al., 1996).
232	Remarkably, Chitral does not receive monsoon rains. That's why it is hard to understand how
233	trees respond to different moisture trends.
234	Here we developed a 467-year (1550-2017) of tree ring chronology of C. deodara, and
235	reconstructed the 424-year (1593-2016) drought variability of the Chitral Hindukush Range in
236	northern Pakistan. The point years, 2002, 2001, 2000, 1999, 1985, 1971, 1962, 1952, 1945, 1921,
237	1917, 1902 and 1892, were recorded in our tree-ring record. Among them, 2001, 1999, 1952, and
238	1921 were identified by previous studies (Khan et al., 2013; Ahmed et al., 2010; Zafar et al.,
239	2010; Esper and Genrt, 2001; Esper et al., 2001; 2002, Ahmed et al., 2009b, 2010a, 2010b; Khan
240	et al., 2008; He et al., 2018). Sigdel and Ikeda (2010) reported that droughts occurred in 1974,
241	1977, 1985, 1993, and the winter of 2001 and the summers of 1977, 1982, 1991, and 1992. Our
242	PDSI reconstruction fully captured the widespread drought in Pakistan, Afghanistan, and
243	Tajikistan in 1970-1971 (Yu et al., 2014). The above drought has disrupted the people's daily
244	life, leading to food and water shortages and livestock losses in high altitude areas (Yadav et al.,
245	2017; Yadav and Bhutiyani, 2013; Yadav, 2011). This drought may also be due to the failure of
246	Western disturbance precipitation (Hoerling et al., 2003).
247	The results were compared with adjacent studies for validation and reliability. Our
248	reconstruction captured a range of climate changes mentioned in other studies (Gaire et al., 2018;
249	Chen et al. 2019; Yu et al., 2014, Ahmad et al., 2004). Our dry periods (1981-1987, 1870-1875
250	to 1761-1764) were consistent with the results of Chen et al. (2019). The dry period 1645-1631
251	also reported in tree-ring based drought variability of Silk Road (Yu et al., 2014). Most notably,
252	three mega drought events in Asian history (Gaire et al., 2018; Yadav, 2013; Panthi et al., 2017),





253	namely, the Strange Parallels drought (1756-1768), the East India drought (1790-1796) and late
254	Victorian Great Drought (1876 to 1878) was clearly recorded in our reconstructed PDSI. Based
255	on the above results, drought changes in Northern Pakistan and central Himalayas are closely
256	related to large-scale ocean-atmospheric circulation and synchronous in western Asia (Gaire et
257	al., 2018). The driest period of 1917-1921 of our reconstruction, followed by another dry period
258	1784-1802, coincided with the eruption of Laki volcano (Iceland) in 1783. This could mean that
259	widespread drought on the continent could be linked to volcanic eruptions (Chen et al., 2019).
260	Interestingly, the wet period of 1995-2016 were very consistent with those of Yadav et al.
261	(2017). These results suggested that the long-term continuous wets in 31 years of the past 576
262	years (1984-2014) may have increased the mass of glaciers in the northwest Himalaya and
263	Karakoram (Cannon et al., 2014; Garg et al., 2014). The devastating floods of July 2010 was also
264	captured by our reconstruction, affecting about a fifth of Pakistan (20 million peoples) (Yaqub et
265	al., 2015). Therefore, we speculate that the size of the Hindukush glacier and the mass of glaciers
266	near to our study areas will continue to increase if it continues to get wet.
267	4.2 The linkage of drought variation with the ocean oscillations
268	The results of wavelet and MTM analysis indicated that the low and high-frequency
269	periods of drought in northern Pakistan may teleconnect with the large-scale climate oscillation.

270 The high frequency of drought cycle (2.1-3.3 years) may be associated with the El Niño-

Southern Oscillation (ENSO) (Van Oldenborgh and Burgers, 2005). The middle-frequency cycle

272 (16 years) may be related to the solar cycle, which is similar with other studies in South Asia

- 273 (Shekhar et al., 2018; Chen et al., 2019; Panthi et al., 2017). Solar activity may affect climate
- 274 fluctuations in the Chitral, Hindukush ranges, northern Pakistan (Gaire et al., 2017). The low-
- 275 frequency (36-38 years) may be caused by the Atlantic Multi-decadal Oscillation (AMO), which





276	is the anomalies of sea surface temperatures (SST) in the North Atlantic basin. Dimri (2006)
277	found that the precipitation surplus in winter from 1958 to 1997 was related to the significant
278	heat loss in the northern Arabian Sea, mainly due to intensification of water vapor flow in the
279	west and the enhancement of evaporation. As a result, large-scale changes in Atlantic
280	temperature could also regulate western Asia's climate. Our result was supported by other
281	dendroclimatic studies (Chen et al., 2019; Sano et al., 2005). The AMO pattern is robust
282	throughout the study area. Precipitation in Mediterranean, Black sea (Alena et al., 2019; Hatwar
283	et al., 2005), and parts of northern Pakistan showed an upward trend from 1980 to 2010, but the
284	precipitation in the Chitral Mountains received from Winter Indian Monsoon (WID) (December-
285	March) and the rainy shadow in summer (Khan et al., 2013). Predicting different climate cycle
286	patterns is not easy. In addition, the flow of UIB depends on changes in the Ablation air mass
287	(Rao et al., 2018; Rashid et al., 2018), so small changes in the Ablation mass may eventually
288	lead to changes in water quality and quantity. The UIB is considered as a water tower in the plain
289	(Immerzeel et al., 2010), so the Hindukush Mountains are particularly important for extending
290	the proxy network and identifying glacier expansion.
291	Our drought reconstruction is very consistent with the point years and drought events
292	recorded in other documents or proxy reconstruction. Due to reconstruction indices, species,
293	geographical differences and other reasons, it is not completely consistent with the whole period.
294	The drought regimes in Hindukush ranges in northern Pakistan may be linked to regional, local,
295	and global climate change. In addition, we only studied the response of C. deodara to different
296	climate change in the Chitral region, northern Pakistan. Therefore, we suggest further
297	dendroclimatic network studies in different areas and species.

298





299 **5** Conclusion

300	Based on the significance of tree-ring width of C. deodara, we developed a 467-years
301	chronology (1550-2017). Considering that EPS threshold is greater than $0.85 (> 5 \text{ trees})$, we
302	reconstructed the current March-August PDSI from 1593 to 2016. Our reconstruction captured
303	different drought changes in different time scale in the Chitral, Hindukush Mountain, Pakistan,
304	which falls in the Indus civilization. Several studies have shown that the Indus civilization
305	changed dramatically in history because of its unpleasant climate. Three historic megadrought
306	events, the Strange Parallel Droughts (1756-1768) the East Indian Drought (1790-1796) and late
307	Victorian Great Drought (1876-1878) were captured by our reconstructed PDSI. These large
308	scale and small droughts may be caused by cold or hot weather. Our results are consistent with
309	other dendroclimatic records, which further supports the feasibility of our reconstruction. In
310	addition, due to a different climate change patterns in the region, we suggested to extend the
311	different proxy networks to understand the remote teleconnection across the continent on the
312	multi-decadal to centennial timescales to meet future climate challenges.
313	

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- 323

324 Data availability

- 325 The reconstructed PDSI can be obtained from the supplementary file of this paper. The tree-ring
- 326 width used in this paper can be download form the International Tree-ring Data Bank.
- 327

328 Author contribution

- 329 Xiaochun Wang and Sarir Ahmad initiated this idea. Sarir Ahmad and Sami Ullah collected
- samples in the field. Liangjun Zhu and Sumaira Yasmeen crossdated and measured the samples.
- 331 Sarir Ahmad and Xiaochun Wang wrote the manuscript. Liangjun Zhu, Yuandong Zhang, and
- 332 Zongshan Li revised the manuscript.

333

334 Competing interest

- The authors declare that they have no conflict of interest.
- 336

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590 Fig captions

- 591 Fig. 1. Map of the weather stations (Chitral station) and sampling sites in the Chitral, HinduKush
- 592 Mountains, Pakistan. The copyright is own to ArcGIS.
- Fig. 2 Monthly maximum, mean, minimum temperature (°C) and total precipitation (mm) in the
 Chitral HinduKush, Mountain, Pakistan (1965-2013).
- Fig. 3 The regional tree-ring width chronology from 1550 to 2017. The gray color represents thesample depth.
- 597 Fig. 4 Person correlation coefficients between tree-ring index of *C. deodara* and monthly total

598 precipitation (1965-2013) and scPDSI (1960-2013) (a) and monthly maximum and

599 minimum temperature (1965-2013) (b) from June of the previous year to September of the

600 current year. Significant correlations (p < 0.05) are denoted by asterisks.

601 Fig. 5 The scPDSI reconstruction in the Chitral HinduKush Mountain, Pakistan. (a) Comparison

between the reconstructed and actual scPDSI; (b) The variation of annual (black solid line)

- and 11-year moving average (red bold line) Mar-Aug scPDSI from 1593 to 2016 with mean
- 604 vale \pm one standard deviation (black dash lines).
- **Fig. 6** Comparison of our PDSI reconstruction (a) with Treydte et al. (2006) precipitation reconstruction (tree-ring σ^{13} C) (b) in Pakistan.
- 607 Fig. 7 The MTM spectrum analysis of the reconstructed scPDSI from 1593 to 2016. Red and
- green line represents the 95% and 99% confidence level, respectively. The figures above thesignificant line represents the significant periods of drought.
- **Fig. 8** (a) Spatial correlation between Actual May-August scPDSI (1901-2017) and Actual May-
- 611 August reconstructed scPDSI (1901-2017). (b) The wavelet analysis of the reconstructed





- scPDSI in the Chitral HinduKush Ranges, Pakistan. The value for p>0.05 were masked out.
- 613 The copyright is own to KNMI.



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Fig. 1 Map of the weather stations (Chitral station) and sampling sites in the Chitral, HinduKush



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- 636 11-year moving average (red bold line) Mar-Aug scPDSI from 1593 to 2016 with mean vale \pm
- 637 one standard deviation (black dash lines).



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August reconstructed scPDSI (1901-2017). (b) The wavelet analysis of the reconstructed scPDSI
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649 in the Chitral HinduKush Ranges, Pakistan. The value for p>0.05 were masked out. The

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⁶⁵⁰ copyright is own to KNMI.





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Table 1. Statistics for a split calibration-verification procedure (p<0.001)

Calibrations	r	R^2	Verification	R^2	RE	CE	ST	DW	RMSE	PMT
1960-2016	0.71	0.51	-	0.50	0.71	0.70	(41, 15)	1.0	0.33	10.0
1989-2016	0.75	0.56	1960-1988	0.55	0.61	0.62	(22, 6)	1.0	0.22	5.80
1960-1988	0.76	0.58	1989-2016	0.56	.636	.623	(19, 9)	0.98	0.22	7.42

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