



- 1 Article
- ² Large-scale, multi-temporal remote sensing of
- ³ palaeo-river networks: a case study from northwest

4 India and its implications for the Indus Civilisation

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13 Abstract: Remote sensing has considerable potential to contribute to the identification and 14 reconstruction of lost hydrological systems and networks. Remote sensing-based reconstructions of 15 palaeo-river networks have commonly employed single or limited time-span imagery, which limits 16 their capacity to identify features in complex and varied landscape contexts. This paper presents a 17 seasonal multi-temporal approach to the detection of palaeo-rivers over large areas based on 18 long-term vegetation dynamics and spectral decomposition techniques. Twenty-eight years of 19 Landsat 5 data, a total of 1711 multi-spectral images, have been bulk processed using Google® Earth 20 Engine Code Editor and cloud computing infrastructure. The use of multi-temporal data has 21 allowed the overcoming of seasonal cultivation patterns and long-term visibility issues related to 22 recent crop selection, extensive irrigation and land-use patterns. The application of this approach 23 on the Sutlej-Yamuna interfluve (northwest India), a core area for the Bronze Age Indus 24 Civilisation, has enabled the reconstruction of an unsuspectedly complex palaeo-river network 25 comprising more than 8000 km of palaeo-channels. It has also enabled the definition of the 26 morphology of these relict courses, which provides insights into the environmental conditions in 27 which they operated. These new data will contribute to a better understanding of the settlement 28 distribution and environmental settings in which this, often considered riverine, civilisation 29 operated.

- 30 Keywords: Multi-temporal; Seasonal; vegetation; palaeo-river; Indus Civilisation; Archaeology
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32 1. Introduction

33 Remote sensing and multi-spectral imagery have considerable potential for the detection of 34 changing river systems and the reconstruction of hydrological networks, particularly those that are 35 no longer easily visible on the ground. Such approaches have been applied to detect relict water 36 courses, palaeo-channels and palaeo-rivers in various regions and environments across the world 37 [e.g. 1-8]. Remote sensing approaches to paleo-hydrology and ancient irrigation display a long 38 tradition and have been particularly fruitful in Mesopotamia and other riverine and 39 channel-dependant Ancient Civilisations. However, most studies have employed single data sources 40 or images acquired at one point in time. While single or limited time-span imagery are a rich source 41 of data, they are unlikely to document annual or long-term variation in sub-surface moisture content 42 and surface vegetation, which are key factors in the detection of palaeo-hydrology [1,8,9].

The promulgation of a wide variety of declassified and low- or no-cost remotely sensedimagery means that a considerable quantity of multi-spectral imagery is now widely available, and

this corpus clearly constitutes big data. The ability to use such large and abundant imagery datasets to their fullest effect is limited, however, by storage capacity and computational power. This paper presents a novel approach to processing big data imagery datasets to detect palaeo-rivers and reconstruct hydrological networks in a specific case-study region in northwest India, which is characterised by both annual and long-term variation in water availability and environmental conditions. The term palaeo-river is used in this paper indistinctively from palaeo-channel to denote a remnant of an inactive river or channel in which the riverbed is filled with sedimentary deposits.

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53 2. Project background

54 The plains of northwest India, which include parts of the states of Punjab, Haryana, and 55 northern Rajasthan are often referred to as the Yamuna-Sutlej interfluve (Fig. 1), as they stretch 56 between the Yamuna and Ganges hydrological system to the east and the Sutlej to the west, which is 57 the easternmost river of Punjab, and part of the greater Indus hydrological system. The flat 58 physiography of the plains of the Indus, Punjab and Yamuna-Sutlej interfluve, in conjunction with 59 their proximity to the Himalayas, and the distribution of winter rain and summer monsoon rain 60 combine to produce a distinctive environmental zone, where water availability is extremely 61 seasonal, has significant variability, and can produce very active fluvial processes [10]. This 62 variability had the potential to have had dramatic consequences for early settlers of the region. Often 63 described as being riverine [e.g. 11], the Indus Civilisation is characterised by a relatively small 64 number of large urban nuclei (e.g. Harappa, Mohenjo-Daro) and an abundance of smaller town and 65 village settlements, which are distributed across the Indus River watershed and its surrounding 66 regions in both modern Pakistan and India. These settlements occur in a wide range of 67 environments, and while a number appear to lie close to watercourses, a significant proportion seem 68 not to have had direct access to perennial water [10,12]. Despite this, there has been considerable 69 historical interest in the hydrology of this region and the plains of northwest South Asia in general, 70 which has been markedly focused on the believed proximity between Indus settlements and relict 71 watercourses [e.g. 13-21].

72 Since the 1980s, satellite images have been employed to trace dry riverbeds and reconstruct the 73 ancient hydrological system in the region [22]. Although these studies have successfully identified 74 multiple palaeo-rivers, they have typically relied on visual interpretation of RGB composites of 75 single multispectral or mosaicked images (mostly Landsat 5-7 multispectral imagery) with visibility 76 being determined by the particular environmental conditions at the moment that the image was 77 acquired. These approaches have largely focussed on the detection of the ancient riverbed of the 78 Ghaggar-Hakra (often identified as the Sarasvati River, which is the only river mentioned in the 79 ancient Vedic texts that does not correspond to any current watercourse) and its most readily visible 80 relic tributaries. While such an approach might be valid for the detection of large palaeo-rivers 81 within a single environmental zone, it only provides limited information on the characteristics of the 82 hydrological system that maintained that river. The reconstruction of as much as possible of a 83 hydrological network is an essential factor to understanding the character and historical 84 morpho-dynamics of any particular river. Environmental changes that could have originated well 85 beyond the visible trace of a riverbed have the potential to affect changing water flow and sediment 86 load, both important factors that could affect the river's morphology, course, and bring about its 87 eventual disappearance.

88 In South Asia, the focus on particular large rivers instead of entire networks might be due to the 89 strong attention given to the relationship between palaeo-channels and the distribution of known 90 archaeological sites, particularly those related to the Indus Civilisation [23-27]. Though not often 91 emphasised, these plains also have evidence for significant occupation in the Early Historic and later 92 periods [28-29], which will also have been affected by hydrological change. In many instances, 93 ancient sites have been used to verify the results of remote sensing or RS-based analysis, and also to 94 construct a chronological framework for the ancient watercourses [e.g. 24–25,30 (pp. 359–384),31], 95 which creates an inherent circularity.



Figure 1. Location of the study area

100 The study area is centred on the Sutlej-Yamuna interfluve (Fig. 1) or the Haryana-Punjab plain, 101 which includes parts of the modern provinces of Haryana, Punjab and Rajasthan in India, and covers 102 an area of approximately 80,000 km². This greater region includes a high diversity of environments, 103 ranging from channel irrigated agricultural areas in the north to the Thar Desert of northern 104 Rajasthan to the south. In geomorphological terms, the area displays a remarkably flat surface with a 105 height difference of 100 m in over 300 km of distance. This plain is comprised of quaternary alluvial 106 deposits originating from the Himalayas [26].

107 This paper has two interrelated research objectives. Firstly, it sets out to demonstrate a robust 108 method for carrying out accurate hydrological modelling of an area with very complex hydrology 109 using large multi-temporal data sets. The second research objective is specifically archaeological. 110 Building upon the work carried out in northwest India by the Land, Water and Settlement project [10], 111 an ERC-funded project entitled 'Winter Rain, Summer Rain: Adaptation, Climate Change, Resilience 112 and the Indus Civilisation (TwoRains)' is exploring the mechanisms involved in the human 113 adaptation to, and management of, variable and changing water availability. In this regard, the 114 reconstruction of the whole palaeo-hydrological network of northwest India will be instrumental to 115 model and test hypotheses on water movement, accumulation, seasonality, and availability in 116 relation to changing climate and rainfall patterns. This paper discusses the first results of an 117 investigation in this direction.

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119 3. Materials and Methods

120 There have been many previous attempts to detect the palaeo-rivers in northwest India using 121 satellite imagery [e.g. 22,32–40]. However, previous RS research has not typically incorporated the 122 whole of the Sutlej-Yamuna interfluve and in the cases where comprehensive RS has been carried 123 out, studies have usually employed single data sources (e.g. Landsat 5-7 multispectral satellite 124 imagery) or several images acquired during the same period, mosaicked over the specific area of 125 interest of each study. These are logical approaches as the reconstruction of the palaeo-hydrological 126 network of a very large area such as the Yamuna-Sutlej interfluve would have been limited until 127 recently to what could be attempted due to limitations in computational power and data availability. 128 Also, this region poses particularly challenging detection problems:

Firstly, the Sutlej-Yamuna interfluve includes a high diversity of environments, from areas of irrigated cultivation to desert. Such variability complicates the visibility of ancient water-related features, as the factors influencing it such as vegetation, seasonal rain, soil moisture and sediment type would vary in areas with different environmental settings. The use of images from a single sensor acquired during the same period would likely be unable to reflect a significant percentage of the features of interest.

135 Secondly, the area has been subject to large-scale landscape modifications and significant 136 land-use changes, particularly since the introduction of mechanised agriculture and extensive 137 irrigation. These inter-related innovations have had an enormous effect in the visibility of 138 palaeo-rivers in multispectral imaging, which was already important in the late 70s as noted by 139 Bakliwal and Sharma [41] (p. 461). Most notably, a large network of irrigation channels (with main 140 distribution branches reaching 50 and 100 m wide) cross the interfluve. These have a double effect in 141 that they: (1) act as an effective barrier to topographic water movement through natural basins, and 142 (2) distribute water to areas that might have not been watered under natural conditions. This 143 interference in the natural hydrology not only diminishes the capacity of ancient courses to retain 144 moisture and increase their visibility, but it also boosts the detection of false positives in the form of 145 relatively modern linear extensions of watered soils served by individual water channels. The 146 introduction and wider access to mechanised agriculture has also facilitated the systematic 147 flattening of large portions of the study area for agricultural purposes, which is a pre-condition for 148 the distribution of channelled water. Although the physiography of the landscape is very flat, 149 features such as ancient habitation mounds and the levees and channels of ancient rivers provide a 150 certain relief, which could have hindered channelled water distribution. The flattening of the 151 landscape resulted in the filling of small palaeo-channels, decreasing their capacity to retain 152 moisture, but also in a strong reflectance of the artificial water channel network and the irrigated 153 fields, which hinders the more subtle indications of ancient watercourses.

Thirdly, seasonal and long-term cultivation changes prevent a uniform visual analysis of the area, which appears as a constantly changing mosaic of cultivated fields in which visibility of palaeo-rivers is partly dependant of the particular cultivation state of the fields planted on the overlying ground surface. This variability again poses a challenge for the use of a single image or images taken during the same period as, at any given moment, only a small part of the area would provide adequate conditions for the detection of palaeo-rivers.

160 Taking into account the environmental particularities of the area of interest and aiming to 161 achieve the maximum detection potential, a large-area, multi-sensor and multi-temporal approach 162 was adopted. This approach aimed to provide a means of overcoming environmental variability, 163 land-use changes, seasonal cultivation patterns and extensive landscape modifications as much as 164 possible. Multi-temporal datasets have the capacity of include diverse environmental and land-use 165 scenarios and their seasonal and statistical treatment can highlight those features that would not be 166 visible during a single particular period, area and/or land-use strategy. In order to implement such 167 an approach Google Earth Engine[©] (EE) was employed as a particularly well-adapted tool to 168 achieve the research aims. EE is a web-based geospatial computing platform with several 169 inter-related components. The Code Editor is an integrated development environment for EE's 170 JavaScript application programming interface. It allows users to write and run their own scripts 171 using EE's JavaScript implementation to import, process, analyse, visualise and export satellite 172 imagery. EE's Code Editor was employed to write and execute the scripts (see supplementary 173 material) necessary to perform the analyses described in this paper. More importantly, EE provides 174 two features that are extremely beneficial for big data multi-temporal remote sensing: (1) access to 175 petabytes of satellite imagery, which includes all Landsat series, ASTER, MODIS, Sentinel 1 and 2, 176 SRTM, and other freely available sources; and (2) capacity to run scripts through Google's cloud 177 parallel computing infrastructure. The former is an important feature as it allows access to more 178 than 40 years of satellite imagery without the need to query and download the data from publicly 179 available repositories, which would have required hundreds of query and download hours and a 180 large storage space. The latter enables the execution of analytical runs using Google's

- 181 high-performance computing cluster; which is essential as the extent of the study area together with 182 the multi-temporal approach adopted by this study, and the size of the necessary multispectral
- 183 dataset is within the realm of Big Data analysis, and well beyond the analytical capabilities of any 184 high-end desktop computer.
- 185
- 186 3.1. Seasonal multi-temporal vegetation indices (SMTVI)
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- January-February mean L5 EVI a. March-April mean L5 EVI Sangrur b. May-June mean L5 EVI c. Samana July-August mean L5 EVI d. Suham September-October mean L5 EVI e. November-December mean L5 EVI 20km f.
- 188 189

190 Figure 2. Seasonal variability of bimensal averages of long-term vegetation indices. Note the higher 191 visibility of palaeo-rivers in the rainy months: January-April (a and b) and July-August (d).

192 193 Preliminary assessment of the study area and the results of previous studies [particularly 22, 194 39-40, but also 36-37] point to a strong relationship between the visibility of palaeo-rivers and the 195 presence/health of surface vegetation, which in turn is related to water availability. Therefore, it was 196 recognised that the use of different indices for the analysis of vegetation health and water content 197 over long periods of time have the potential to offer important insights into the tracing of 198 palaeo-rivers and channels. In order to explore this relationship, different vegetation indices derived 199 from Landsat 5 imagery were employed. Despite its reduced spectral resolution with respect to later 200 satellites, Landsat 5 was considered the best source to develop this study as it has produced almost 201 30 years of imagery (from 1984 to 2013), some of it acquired prior to the construction of some of the 202 large water channels in the area. It therefore provides an ideal source for developing long-term 203 multi-temporal imagery analysis. The EE image repository provides NDVI, EVI [42] and NDWI [43] 204 annual, 32 and 8-day composites of Landsat 5 imagery. For the study area, their calculation included 205 2266 Landsat 5 TM level 1T orthorectified scenes, using the computed TOA reflectance [44] and 206 filtered from cloud cover [45]. These were acquired from Jan 1, 1984 to May 8, 2012. A script was 207 written (available as supplementary material) to group the 1254 available 8-day EVI composites into 208 single mean two-month images. Each of these was the average of all 8-day EVI composites for the 209 same two months from 1984 to 2012 for the study area. It was decided to select 8-day instead of 210 32-day composites in order to preserve the temporal integrity of the data as much as possible while 211 the grouping of the vegetation indices into two-month averaged values would allow exploring 212 seasonal variability. Every multiyear bimensal image was the result of averaging around 200 8-day 213 composite images.

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- 215
- **Figure 3.** Image resulting of the application of SMTVI for the rainy months as an RGB Composite of
- 217 the 1984-2012 EVI averages (R: Jul-Aug, G: Mar-Apr, B: Jan-Feb).

219 The images resulting from the application of the different vegetation indices show similar 220 results, with EVI producing a slightly clearer identification of palaeo-rivers (Fig. 2). Seasonality 221 seems to be a much more important factor than the type of vegetation index employed, as the images 222 from January-February, March-April and July-August present a surprisingly clear visibility of 223 palaeo-rivers, which are undetectable in other images (Fig. 2). This seasonal difference is probably 224 related to the fact that these months closely correspond to the Indian rainy seasons. Two methods 225 were tested to facilitate the visualisation of the entire palaeoriver network: a RGB composite of the 226 three principal components from a PCA of all bimensal images; and a Seasonal Multi-Temporal 227 Vegetation Index (SMTVI), which resulted from creating a RGB composite of the three EVI images 228 corresponding to the Indian Winter rainy season and Summer Monsoon months (January-February, 229 March-April and July-August). Although most palaeo-rivers were visible in the principal 230 component composite, SMTVI (Fig. 3) offered a better visibility, probably related to the reduction in 231 non-relevant information (i.e. months in which palaeo-rivers were not visible).

In order to further explore the seasonal variability of the vegetation indices a normaliseddifference vegetation seasonality index (NDVSI) was developed using the following calculation:

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$\frac{((\bar{x}WNIR + \bar{x}WRED) / \bar{x}WNIR) - ((\bar{x}DNIR + \bar{x}DRED) / \bar{x}DNIR)}{((\bar{x}WNIR + \bar{x}WRED) / \bar{x}WNIR) + ((\bar{x}DNIR + \bar{x}DRED) / \bar{x}DNIR)}$

235

236 Here xW represents the mean values of L5 images during the rainy seasons and xD the mean 237 values of L5 images during dry months. The NDVSI image (Fig. 4) provided a good guide on the 238 areas where seasonal vegetation variability was at its greatest and, therefore, the use of seasonal 239 vegetation indices had the potential to yield a higher amount of information. An immediate result of 240 the application of the NDVSI was the clear difference between the northern and southern sectors of 241 the study area (Fig. 4). While the northern sector has the potential to provide new information based 242 on the seasonal variability of vegetation, the exploration of seasonal variability of vegetation indices 243 could barely produce relevant results in the southern sector. This is probably due the fact that this 244 area has a hotter and dryer climate and is currently dependent on irrigation for crop cultivation. 245



246

247 Figure 4. Normalised Difference Vegetation Seasonality Index (NDVSI) of the study area where the

248 greenest areas indicate a higher seasonal variation of vegetation. Ghaggar-Hakra palaeo-channel has

251 *3.2. Spectral decomposition techniques*

252 Based on the indications provided by the NDVSI, a different approach making use of spectral 253 decomposition techniques, in particular Tasselled Cap Transformation (TCT) and Principal 254 Component Analysis (PCA) of multispectral imagery, was also applied in the study area. These were 255 considered good exploratory techniques as they both reduce the amount of information available in 256 the multispectral datasets to weighted sums of separate channel readings. These techniques effectively reduce data dimensionality facilitating their interpretation. Both TCT [46-48] and PCA 257 258 were produced using the mean values of all Landsat 5 images available for the study area. In 259 addition to this, and in order to test the seasonal influence in the visibility of palaeo-rivers, TCTs and 260 PCAs were calculated for the mean of images acquired during the Indian winter rain season and 261 summer Monsoons and during the dry months.

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The combination of these different approaches was designed to increase the detection of palaeo-rivers and decrease the occurrence of false positives. Figure 5 provides a description of the workflow followed. Figure 5 also includes indications on the preferential application of the different techniques following the guide provided by the NDVSI. A detailed account of the results obtained follows.

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Figure 5. Workflow followed for the generation of the different outputs described in the text andtheir recommended application according to seasonal vegetation variability.

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274 4. Results

A total sum of 8246 kilometres of palaeochannels was mapped using both SMTVI and spectral decomposition techniques (Fig. 6). The palaeo-rivers and channels were delineated by hand using a vector-based GIS desktop platform as testing of edge/line detection algorithms and automatic processing did not provide reliable enough results. The resulting map of palaeo-rivers in the area cannot in any way be taken as definitive, as there might have been multiple palaeo-rivers that remain undetected. This is particularly true of the southernmost sector of the study area (south of

- the main Ghaggar-Hakra channel) as defined by the results of the NDVSI analysis (Fig. 4). The
- results of the two methods adopted will now be described and analysed.



Figure 6. Results of the interpretation of both SMTVI and seasonal multitemporal spectral decomposition
 techniques for the reconstruction of the palaeo-hydrological network of the Sutlej-Yamuna interfluve.

287 4.1. SMTVI

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SMTVI resulted in the clear visualisation of a range of formerly reported palaeo-rivers, but also multiple previously undetected ones. This approach only failed to yield results in the southern sector of the area under study, in particular below the main Ghaggar-Hakra channel where dune morphologies dominate the landscape. In this area, the results of the NDVSI analysis (Fig. 4) indicate a very low seasonal variability of vegetation.

293 A total of 5034 kilometres of palaeo-rivers have been mapped through SMTVI, and these were 294 predominantly in the northern sector of the study area (Fig. 6). The results of this technique show 295 that the traces of palaeo-rivers form a more complex network than previously suspected. Previous 296 studies in this area have been successful in detecting a number of these palaeo-rivers (Fig. 7). For 297 example, Yashpal et al. [22] and van Dijk et al. [40] results are partly coincident with our own even if, 298 in the Yashpal et al. [22] case, some of these have been joined to form continuous channels for which 299 we have not found evidence. In several instances, palaeo-rivers detected by these authors seem to be 300 coincident with large modern irrigation channels, which is unsurprising given the early date and 301 small scale at which the analysis was conducted. SMTVI has also been unable to find traces of the old 302 channel of the Sutlej and the old channel of the Yamuna 1, as described by Yashpal et al. [22]. All the 303 features detected in the northern sector of the study area by van Dijk et al. [40] have been identified, 304 as have most of those detected in the west sector, but we have also detected many others not documented by van Dijk *et al.* [40] and Bhadra *et al.*'s [39] hypothesised location of palaeo-rivers in the area do not coincide with the results of our study nor with any other previous analyses under consideration. In general, it can be said the SMTVI provided a more consistent approach for the medium resolution detection of palaeo-rivers across a very large area displaying higher values of seasonal vegetation variability.





Figure 7. Palaeo-rivers detected in the study area by previous RS-based studies.

312 The visibility of the palaeo-rivers in SMTVI (Fig. 3) is a reflection of the interplay between past 313 fluvial geomorphology and current multi-year vegetation dynamics. In this regard, a more detailed 314 analysis of the characteristics of the detected features can help the understanding of past 315 environmental conditions. Monsoonal rains imply a strong rise of soil moisture and surface 316 vegetation. However, palaeo-rivers in SMTVI are not detected through higher vegetation values 317 but, on the contrary, as linear groupings of low value pixels, that is, as features where the average 318 vegetation values during the last thirty years have been significantly lower than in their 319 surroundings. The very low terrain slope (average of 0.37%) and the high volume of rainfall during 320 the winter and summer rain periods, which merge and flow down the foothills of the Himalayas, 321 combine to create a large depositional floodplain. The rivers that have been documented form a 322 parallel drainage pattern and present a highly sinuous and meandering morphology in which 323 multiple meanders and meander scars are visible. The low value pixels that define their trace seem 324 to correspond to two types of features: (1) large natural levees typically formed after multiple 325 flooding episodes [49]. These are usually elevated with respect to the floodplain, as confirmed by 326 SRTM 30m data, and accumulate thicker and coarser sediment [50], and (2) coarser channel

327 deposits and alluvial deposits accumulated in the inside of river bends. Both proximal areas of 328 levees, inner river bends and channel bedload tend to concentrate coarser sediments [51], which 329 might have resulted in less fertile soils with higher water subsurface infiltration and lateral water 330 runoff in the case of the levees. Water would have been concentrated and retained by the finer 331 material and flatter topography present at the backswamp areas of the floodplain, where bright 332 pixels reflect a correlation between seasonal rainfall and healthier vegetation. The contrasting 333 vegetation response between the coarser deposits close to the river channel and their surrounding 334 floodplain account for the clear palaeo-river visibility in this area. The presence of ribbon-like 335 deposits and their attendant levees is consistent with the existence of rivers with a high suspended 336 sediment fraction forming a rapidly aggrading system where avulsive channels are common [51] (p. 337 6).

338 Several relevant results for the reconstruction of the hydrologic history of the northern sector of 339 the study area have been obtained through the use of seasonal vegetation mapping: (1) the 340 confirmation of a major palaeo-course of the later Sutlej river, which contributed to the 341 Ghaggar-Hakra system, though when and for how long remains unknown (top right corner of the 342 lower image in Fig. 3,); (2) the migration of this same major watercourse from the Ghaggar-Hakra 343 catchment to that of the Sutlej, which would have significantly reduced the amount of water 344 available in the Ghaggar-Hakra's lower course; and perhaps most significantly; (3) the 345 multiplication of the palaeo-rivers known in the area, which indicates that as a whole, the region has 346 an extremely complex fluvial history, which will have had important and as yet poorly resolved 347 consequences for water availability and thus also for past human habitation and land-use. SMTVI 348 also allowed study of the morphology of the palaeo-rivers and documentation of multiple avulsion 349 episodes, with consequences for the human habitation and use of the area through which these 350 flowed.

The use of vegetation indices, however, did not offer any indication of the location of palaeo-rivers in the southeastern sector of the study area. This area is characterised by arid conditions and a reduced rainfall and, as NDVSI (Fig. 4) clearly shows, seasonal variability does not produce significant changes in vegetation. Vegetation indices are, therefore, not considered a consistently reliable approach for the detection of palaeo-rivers in this area.

356 *4.2. Spectral decomposition techniques*

357 A total of 1920 kilometres of palaeo-channels have been detected through the use of spectral 358 decomposition techniques of seasonal multi-temporal data. Not surprisingly, TCTs of the mean 359 Monsoonal image composite values for Greenness and Wetness produced similar results for the 360 northern area to those obtained with the application of vegetation indices. The RGB composite of the 361 different bimensal EVIs was more efficient in detecting palaeo-rivers due to the higher temporal 362 resolution of the data employed, which incorporated differences in seasonal rains. In contrast to the 363 vegetation indices, TCT, however, excelled in the detection of palaeo-rivers in the southern sector of 364 the study area where a large relic water course with several tributaries contributing to the 365 Ghaggar-Hakra system was identified and its trace could be followed for more than 300 kilometres 366 (Fig. 8). This feature was visible through its contrasting low value pixels in the fourth (where it was

367 clearly discernible), sixth and only marginally in the second axis corresponding to greenness. The 368 palaeo-river was noticeably more visible in the TCT of the dry months composite (Fig. 8), slightly 369 less in the TCT of the mean image of all available Landsat 5 imagery and scarcely visible in TCT of 370 the monsoonal composite. This pattern of visibility indicates that, although seasonally variable, the 371 visibility of this palaeo-river is not related to vegetation changes. This palaeo-channel might have 372 been previously documented in part and with large inaccuracies by Yashpal et al. [22], who 373 considered it a second old channel of the Yamuna River (fig. 7). Later Rajani and Rajawat [37] 374 identified its terminal sector as the Vedic Drishadvati. Recently, Mehdi et al. [39] again identified this 375 palaeochannel and reliably traced its course and those of several tributaries covering a total length of 376 approximately 416 km.



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Figure 8. Upper image: RGB (TCT axes 1, 4 and 2) composite of a multi-temporal seasonal mean of
dry months Tasselled Cap Transformation. Note the higher albedo in the zoomed areas (images A
and B, Sentinel 2 RGB natural colour composite of bands 4, 3 and 2) from Nohar to Bhadra.

381 Although slightly meandering, the course of this river does not show evidence of levees or 382 aggradation, neither are these features present in the 30 m/cell SRTM data. Instead, its course is 383 marked by a clear erosive channel, which corresponds to a higher slope than that documented from 384 the northern plain. The orientation of this palaeo-river in conjunction to topographic data suggests 385 that it might have been collecting water from different sources. While its northern section could have 386 been sourced from the Yamuna River catchment, as suggested by Yashpal et al. [22], its southern 387 tributary, which joins the northern channel in Bhadra (Fig. 8b), appears to originate in the Aravalli 388 range. This area of the Aravalli has an important component of carbonates. Dissolved carbonates 389 could have been incorporated in the water and influenced the composition of the sediments 390 transported by the river and deposited along its banks. Sediment rich in calcium carbonate has a 391 clear colour, which could be responsible for the noticeably higher albedo of the fields along the 392 palaeoriver course (Fig. 8). The presence of carbonate, gypsiferous clay and strings of playa lakes are 393 all considered residual elements of palaeo-channels [9] (p. 49). Although, little geological coring has 394 been done in this area, the presence of gypsum and carbonates is well documented in Karsandi 395 palaeo-environmental record [52-53], twenty-one kilometres south of this channel at Nohar. This 396 particular composition of the channel's filling sediment could also explain the decreased visibility of 397 the palaeo-river during the rainy season, as higher soil moisture would have reduced the contrast of 398 carbonate-rich sediment. SRTM data also show how the southernmost tributary of this river is 399 completely obliterated under the dunes marking the northeastern edge of the Thar Desert (Fig. 8, 400 dunes in bright yellow).

401 The southeastern sector of the study area between Hisar and the Yamuna (Fig. 9), where NDVSI 402 analysis produced average values for seasonal vegetation variability, also offered interesting results. 403 Figure 9a shows a strong visibility of palaeo-rivers and channels in the southeastern extreme of the 404 study area only during wet months, where an anastomosing river with multiple meander scars can 405 be appreciated coming from the Yamuna bluff line. TCT of images acquired during the wet months 406 shows similar detection capabilities to that of SMTVI (Fig. 9b). In contrast, the complex network of 407 palaeochannels west of Hisar, and much closer to the Thar Desert edge, identified in Figure 9c area 408 was only visible in the TCT produced from images taken during dry months. In many cases, 409 particularly for the spectral decomposition techniques and the southeast sector, the alternation of 410 multi-temporal seasonal means for dry and wet seasons provided complementary identifications as 411 some palaeo-rivers or stretches of these were only visible during dry or wet seasons. The 412 combination of seasonal images was also important beacuse it allowed a clear differentiation 413 between palaeo-rivers and irrigated fields, which formed linear and dendritic patterns (similar to 414 those typical of palaeo-rivers) following water channels. Artificial channel-irrigated fields were 415 particularly visible in the composites of dry seasons as their moisture content boosted their visibility 416 with respect to their surrounding fields. Their identification was an important step to avoid the 417 mapping of false-positive palaeo-rivers. This clearly exemplifies the importance of seasonality for 418 the identification of palaeo-features in those areas where NDVSI shows average values for seasonal 419 vegetation variability.

420 Seasonal PCA analysis of the mean values of L5 images produced similar, if less clear, results to 421 those of the TCT. The northern network of palaeo-rivers was visible, although with much less detail 422 than that obtained with the seasonal EVI analysis. This indicates the close relationship between the 423 visibility of northern palaeo-rivers and long-term vegetation dynamics, which diffuses with the use 424 of spectral decomposition techniques. The southern palaeoriver network (Fig. 8) was only visible in 425 the dry season PCA, where it displayed high values in principal components 3 and mostly 4, where 426 it was clearly identifiable. It seems therefore that seasonal PCA could provide a good approach to 427 areas with less vegetation. Also the visibility of this palaeo-river network in the more marginal axes 428 of TCT and PCA points to the inadequacy of the more conventional RS techniques usually applied in 429 the detection of the ancient rivers in the Indus watershed. The success of the approach adopted by

- 430 Mehdi *et al.* [39] in detecting a channel (Fig. 9) joining the modern town of Hisar and the Indus city of
- 431 Rakhigarhi is probably due to specific environmental conditions during the acquisition of the single
- 432 Landsat 7 image from which these authors developed a PCA. Unfortunately, no information about
- this particular image was provided, and it thus not possible to evaluate their reconstruction.
- 434



Figure 9. Comparison of visibility potential between different seasonal conditions and techniquesillustrating the need to combine different approaches for the detection of palaeo-rivers.

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For those areas with average and low NDVSI values PCA seasonal analysis provided a
complementary source of information to previous TCT and SMTVI data and was useful to increase
the number of detected traces while reducing the number of false positives.

443 5. Discussion

444 The analysis presented here demonstrates that the palaeo-rivers in the study area can take 445 multiple forms as a result of topographic and sedimentologic factors in conjunction with discharge 446 and hydrological parameters. Today, both smaller palaeo-channels and medium-sized past fluvial 447 courses have been almost completely filled in and covered by a flattened and cultivated land surface. 448 The largest palaeo-channels can also be preserved as subtle topographic features, which range from 449 accumulated levee sediments and flattened filled up channels to carved channels. Sometimes these 450 stages can all be related to different reaches of the same river as slope varies and varying amounts of 451 water and sediment are incorporated.

The morphology, topography, and multiplicity of palaeo-rivers in the northern sector of the study area together with their spectral response to vegetation indicates that these were very active 454 aggradational environments in which rivers carried a high suspended sediment fraction. Floods 455 were a common feature of this landscape as suggested by the strong development of river levees. Flooding episodes in conjunction with the high bedload would have strongly contributed to the 456 457 avulsion of many of these palaeo-rivers at one or several stages of their course. The agricultural use 458 of these floodplains would have required good adaptation to flooding regimes and changing water 459 conditions (including the specific location of water courses). The adoption of elevated settings for 460 habitation and mobility are common past human responses to this type of variable environmental 461 conditions [10,54–55]. It seems unlikely that settlements with a strictly agricultural economic 462 orientation would have been situated on the proximal areas of these rivers' levees, not just because 463 of the risk of flooding or river avulsion, but also of the lower agricultural productivity of these areas. 464 The more distal areas of the levees, backswamps and floodplain would have offered better soils 465 (more organic and with finer sediment) and a higher capacity for moisture retention. In this regard, 466 SMTVI offers not just a way to detect palaeo-rivers but it can also be employed to explore 467 agricultural potential.

468 The southern sector of the study area indicates an important influence of changing climatic 469 conditions linked to the expansion of the Thar Desert and the progressive reduction and eventual cut 470 off of water coming from the Aravalli range. The expansion of the Thar Desert toward the north 471 must be accounted for, not only because of the eventual disappearance of this channel but also 472 because of the possible presence of a more extended network of rivers coming from the Aravalli 473 range and hidden beyond the capacity of current RS techniques under the expanding dunes of the 474 Thar Desert. Singhvi and Kar's [56] review of the evidence for dune formation in the Thar Desert 475 noted a west to east expansion during the mid-Holocene. This sector of the study area is located in 476 the northeastern extreme of the Thar Desert with a similar rainfall pattern and latitude to that of 477 Lake Didwana, which dried at around 4200 BP [57]. This increase in aridity, noted in the Didwana 478 record could have reduced the amount of water available in these palaeo-rivers but, most 479 importantly, marked the beginning of an important dune development period in the north-eastern 480 margin of the Thar desert, which might have been coincident with the Post-Urban or Late Harappan 481 phase starting around 3900 BP (c.1900 BC). Although, dune formation in the Thar desert is a long 482 process spanning at least 200ka BP [56], some recent dating of dunes from the northernmost edge of 483 the Thar desert, have suggested that dune deposits pre-date 4.9 ka BP, though some are around 1500 484 years old [27 (p. 4),58–60]. Therefore, it seems logical to expect a long period of dune formation and 485 interaction with rivers before these were finally erased. Interdunal areas at the desert edges are 486 important reservoirs of surface and subsurface water during inter-monsoonal periods and have been 487 long inhabited and exploited as highlighted by research at Gujarat [61]. The reconstruction of a large 488 palaeochannel in this area flanked by dunes and with a tributary emerging from a dune-dominated 489 area (dunes are shown in yellow in Fig. 8) might have been an important resource for the area's 490 inhabitants even after dunes started to expand in this region.

491 The increase in aridity and the related weakening of the Summer Monsoon during the 4.2 ka BP 492 event, which may have been linked to the extension of the Thar Desert, might have also been a factor 493 in the particular fluvial morphology detected in the northern sector of the study area. The natural 494 increase in rivers sediment load has been employed as a proxy for dryer climatic conditions [e.g. 62]. 495 In this regard, it is tempting to correlate the northern rivers' large ribbon shaped levees and avulsion 496 episodes, closely related with high sediment loads, with a dryer period. If this were the case, the 497 migration and avulsion towards the Sutlej catchment of the northern rivers and the subsidence of the 498 southern network through the extension of dunes would have combined to significantly reduce the 499 amount of water available to the Ghaggar-Hakra. However, the current lack of chronological data on 500 the periods of activity of the detected rivers prevents us from investigating such hypotheses yet.

501 The data provided by these analyses are also important in contextualising previous studies in 502 which palaeo-rivers have been dated using the distribution of known archaeological sites. 503 Notwithstanding the positional accuracy of these locations [see 10,63–64], which would severely 504 hamper their use for validation purposes, the results for the northern sector of the study area (Fig. 3) 505 suggest that proximity to the river might not be a good indication of contemporaneity as the fields 506 close to the river channel might not have been the most productive in agricultural terms. Sites with 507 an agricultural orientation might have preferred to occupy elevations above flooding level in the 508 finer sediment accumulation area. Flooding events and changing river courses could also have had 509 an important effect in the preservation of archaeological sites eroding and burying those that were 510 located in the path of new courses or in their sediment accumulation areas.

511 In addition to these factors, the scale at which these correlations between specific 512 palaeo-channels and settlement locations have usually been published [e.g. 30 (pp. 359–387), 31 (Fig. 513 4.2)] do not allow the accurate correlation between the two elements. The use of large area (small 514 scale in a geographic sense) site distribution maps for these correlations results in a visual 515 association between the shape of the palaeo-river and the line formed by the grouping of sites, but at 516 these scales the sites could be aligned to a number of palaeo-channels given the number of rivers and 517 the parallel morphology of the drainage in the Sutlej-Yamuna interfluve revealed here. This analysis 518 thus suggests that at these scales it is not possible to co-relate lineal distribution of archaeological 519 sites to any particular palaeo-river that we have documented. The results from previous studies 520 reconstructing the chronology of the hydrological system using the position of archaeological sites 521 and vice versa [e.g. 24–25,30 (pp. 359–384),31] are, therefore, considered unreliable.

522 The adoption of a multi-technique seasonal approach (SMTVI, Seasonal TCT and seasonal PCA) 523 has allowed avoiding false positive identifications while achieving a high degree of accuracy and 524 multiplying the number of known palaeo-rivers as demonstrated by the comparison with previous 525 studies carried out in the study area. However, since the current knowledge of site distribution 526 patterns and the nature of their relationship to ancient river courses are inaccurate at best 527 quantification of error, which has usually employed the distribution of known archaeological sites, 528 is difficult to measure at this stage. Historic map analysis and fieldwork aimed at the verification of 529 site locations and palaeo-rivers is currently being developed and will be the object of future 530 publications.

531 6. Conclusions

While previous remote sensing approaches using single images acquired at the moment of maximum visibility of a particular palaeoriver can offer excellent visibility of specific features, they will not necessarily be (and often simply will not) be able to reveal other aspects of a hydrologic network. Of course, these approaches would be unviable for the reconstruction of the entirety of a complex network, which in this case includes a very large area that can only be covered by hundreds of images.

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539 For this study more than 8000 kms of palaeo-rivers have been mapped and, although a part of 540 those have been already identified by previous studies, our results show that previous approaches to 541 the study of the river systems of the Yamuna-Sutlej Interfluve need major reconsideration. This 542 study thus has important implications for perceived spatial correlations between the distribution of 543 archaeological sites and the course of particular palaeo-rivers. RS visibility patterns have provided 544 important insights into the fluvial geomorphology, water distribution and depositional history of 545 the study area, which, in turn provide significant new data to model the human occupation of these 546 spaces. Given the complexity of the hydrological system, the variety in the climatic and weather 547 system of this region, and the diversity of ways that ancient populations are likely to have obtained 548 water, it is unwise to use the date of occupation at specific settlements to date when specific channels 549 carried water. It is essential to date the different palaeo-courses independently to properly 550 reconstruct the evolution of hydrological networks over long periods. The chronologically consistent 551 reconstruction of this palaeo-river network would allow the testing of different hypothetical

scenarios of water availability through the use of network analysis in combination to hydrological analysis. Future work within the *TwoRains* project will concentrate in: (1) the gathering of accurate palaeo-environmental and cultural data and absolute dating evidence to date the activity of the different palaeo-rivers detected; and (2) creating a database of well-dated and accurately positioned archaeological sites that could allow the statistical inference of relationships between distribution of sites and palaeo-rivers and their potential flooding areas.

558

559 The important increase in the detected palaeo-rivers achieved by this study can be attributed to 560 the use of multi-temporal data filtered by season. Previous studies using single data sources or a 561 selection with scarce temporal variability have failed to factor the multiple environmental and 562 cultural (cultivation patterns) conditions that determine the visibility of palaeo-rivers. The use of a 563 seasonal multi-temporal approach has allowed us to overcome seasonal visibility problems 564 associated with soil moisture and vegetation health, yearly changes in cultivation patterns and 565 long-term changes in the selection of cultivated crops. In this regard, this study clearly illustrates the 566 need to employ a multitemporal seasonal approach and a combination of data treatment techniques. 567

568 Our results prove that the factors influencing water availability along the Ghaggar-Hakra basin 569 are much more complex than previously thought. The traces of palaeo-rivers that have been 570 identified cover the entirety of the landscape in the northern sector forming an almost continuous 571 parallel pattern, which points to the changing nature of these channels and the likelihood that floods 572 and river avulsions have been a relative common occurrence. The waters feeding the various 573 palaeo-rivers originated from glacier-fed sources, such as water supplying the various palaeo-rivers 574 related to the Sutlej, which appear to include the main Ghaggar-Hakra channel, as well as 575 monsoonal rain which is likely to have contributed to both perennial and ephemeral rivers [see 10, 576 27,64]. The geographic source of watercourses ranges from the Himalayas to the Aravalli mountains, 577 and seasonal rain patterns and discharge across this zone are very different. All these factors join to 578 create an extremely complex picture in which water availability and location is dependent upon a 579 multiplicity of factors and difficult to predict in the long term.

580

The methodology outlined in this paper and its associated code in combination with the resources made available through Google[®] Earth Engine [66] can easily be applied to other areas with similar problematics. Given the importance of water sources and water availability for the flourishing, expansion and environmental resilience of ancient civilisations, the development of new methods for the reconstruction of ancient waterscapes is of great importance. Our results highlight the enormous potential of satellite-based Big Data analysis for archaeological research, which, with notable exceptions [e.g. 67], are still underexploited.

588

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600

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605 References

- Bisson, M.; Piccinini, S.; Zanchetta, G. A Multidisciplinary GIS-Based Approach for Mapping Paleoriver Migration: A Case Study of the Serchio River (Lucca Alluvial Plain, Tuscany). *GISci Remote* Sens 2011, 48(4), 566–582.
 http://www.tandfonline.com/doi/abs/10.2747/1548-1603.48.4.566 (accessed on 02/06/2017).
- Blumberg, D.G.; Neta, T.; Margalit, N.; Lazar, M.; Freilikher, V. Mapping exposed and buried drainage systems using remote sensing in the Negev desert, Israel. *Geomorphology* 2004, *61*, 239–250, 10.1016/j.geomorph.2003.12.008. <u>https://doi.org/10.1016/j.geomorph.2003.12.008</u>
 (accessed on 02/06/2017).
- McCauely, J.F.; Blom, R.; Breed, C.S.; Elachi, C.; Grolier, M.J.; Haynes, C.V.; Issawi, B.; Schaber,
 G.G. Subsurface valleys and geoarchaeology of the Eastern Sahara revealed by shuttle radar. *Science* 1982, 218, 1004–1020, 10.1126/science.218.4576.1004.
 http://science.sciencemag.org/content/218/4576/1004 (accessed on 02/06/2017).
- 618 4. Orengo, H.A.; Ejarque, A.; Albiach, R. Water management and land-use practices from the
 619 Iron-Age to the Roman period in Eastern Iberia. *J Archaeol Sci* 2014, 49, 265–275,
 620 10.1016/j.jas.2014.05.005. <u>http://www.sciencedirect.com/science/article/pii/S0305440314001782</u>
 621 (accessed on 02/06/2017).
- 622 5. Rossetti, D.F.; Góes, A.M. Late Quaternary drainage dynamics in northern Brazil based on the
 623 study of a large paleochannel from southwestern Marajó Island. *An Acad Bras Cienc* 2008, *80*(3),
 624 579–593, 10.1590/S0001-37652008000300017.
 625 <u>http://www.scielo.br/scielo.php?script=sci_arttext&pid=S0001-37652008000300017</u> (accessed on
 626 02/06/2017).
- 6. Tapley, I.J. The Reconstruction of Palaeodrainage and Regional Geologic Structures in Australia's Canning and Officer Basins Using NOAA-AVHRR Satellite Imagery. *Earth-Science Reviews* 1988, 25, 409–425, 10.1016/0012-8252(88)90008-6.
 http://www.sciencedirect.com/science/article/pii/0012825288900086 (accessed on 02/06/2017).
- 631 7. Walstra, J.; Heyvaert, V.M.A.; Verkinderen, P. 2009 Remote sensing for the study of fluvial
 632 landscapes in Lower Khuzestan, SW Iran. In Remote Sensing and Photogrammetry Society
 633 Conference, Leicester, UK, 8-11th September 2009.
- 8. Yang, X.; Damen, M.C.J.; van Zuidam, R.A. Satellite remote sensing and GIS for the analysis of
 channel migration changes in the active Yellow River Delta, China. *INT J Appl Earth Obs* 1999,
 1(2), 146–157, 10.1016/S0303-2434(99)85007-7.
 http://www.sciencedirect.com/science/article/pii/S0303243499850077 (accessed on 02/06/2017).
- 638 9. Hou, B.; Mauger, A. How well does remote sensing aid palaeochannel identification? an
 639 example from the Harris Greenstone Belt. *MESA Journal* 2005, *38*, 46–52.

Remote Sens. 2017, 9, x FOR PEER REVIEW

- https://sarigbasis.pir.sa.gov.au/WebtopEw/ws/samref/sarig1/image/DDD/MESAJ038046-052.p
 df (accessed on 02/06/2017).
- Petrie, C.A.; Singh, R.N.; Bates, J.; Dixit, Y.; French, C.A.I; Hodell, D.; Jones, P.J.; Lancelotti, C.;
 Lynam, F.; Neogi, S.; Pandey, A.K.; Parikh, D.; Pawar, V.; Redhouse, D.I.; Singh, D.P.
 Adaptation to variable environments, resilience to climate change: investigating Land, Water
 and Settlement in northwest India. *Curr Anthropol* 2017, *58*(1), 1–30, 10.1086/690112.
 http://www.journals.uchicago.edu/doi/full/10.1086/690112 (accessed on 02/06/2017).
- 647 11. Marshall, J. *Mohenjo-daro and the Indus Civilisation;* Arthur Probsthain: London, UK, 1931.
- 648 12. Wright, R.P. *The ancient Indus. Urbanism, economy and society;* Cambridge University Press: New
 649 York, USA, 2010.
- 13. Oldham, C.F. Notes on the Lost River of the Indian Desert. *Calcutta Review* 1874, 59, 1–29.
- 14. Oldham, C.F. The Sarasvati and the Lost River of the Indian Desert. *J Roy Asiatic Soc* 1893, 34, 49–76, 10.1017/S0035869X00022176.
 https://www.cambridge.org/core/journals/journal-of-the-royal-asiatic-society/article/art-iiithe-s araswati-and-the-lost-river-of-the-indian-desert/D89CD30F203FE92F1EF0AD9B64748A4F
 (accessed on 02/06/2017).
- 656 15. Oldham, R.D. On probable changes in the geography of the Punjab and its rivers: an
 657 historicogeographical study. *Journal of the Asiatic Society of Bengal* 1886, 55, 322–343.
- 658 16. Stein, M.A. A survey of ancient sites along the lost Saraswati river. *Geogr J* 1942, *99*, 173–182,
 659 10.2307/1788862. <u>http://www.jstor.org/stable/1788862</u> (accessed on 02/06/2017).
- 660 17. Ghose, A. The Rajputana Desert Its Archaeological Aspect. *Bulletin of the National Institute of*661 *Sciences of India* 1952, *I*, 37–42.
- 18. Lambrick, H.T. *Sind: a General Introduction*. History of Sind Series 1, Sindhi Abadi Board:
 Hyderabad, India, 1964.
- Lambrick, H.T. The Indus flood-plain and the "Indus" civilization. *Geogr J* 1967, 133, 483–494,
 10.2307/1794477. <u>http://www.jstor.org/stable/1794477</u> (accessed on 02/06/2017).
- 20. Raikes, R. Kalibangan: Death from Natural Causes. *Antiquity* 1968, 42, 286–291,
 10.1017/S0003598X00034505.
 https://www.cambridge.org/core/journals/antiquity/article/kalibangan-death-from-natural-cau
 ses/7EDD56BFE2093C303537BEDD3CF9E090 (accessed on 02/06/2017).
- 670 21. Rajaram, N.S. Sarasvati River and the Vedic Civilization: History, Science, and Politics. Aditya
 671 Prakashan: New Delhi, India, 2006.

- 672 22. Yashpal, S.B.; Sood, R.K.; Agarwal, D.P. Remote sensing of the 'Lost' Sarasvati River. *P Indian*673 *AS-Earth* 1980, *89*(3), 317–331.
- 674 23. Mughal, M.R. Ancient Cholistan: Archaeology and Architecture. Ferozsons: Lahore, Pakistan, 1997.
- 24. Lal, B.B. *The Sarasvati flows on: the continuity of Indian culture*. Aryan Books International: NewDelhi, India, 2002.
- 677 25. Valdiya, K.S. *Saraswati: the River that Disappeared*. Indian Space Research Organisation and
 678 Universities Press: Hyderabad, India, 2002.
- 679 26. Srivastava, G.S.; Singh, I.B.; Kulshrestha, A.K. Late quaternary geomorphic evolution of
 680 Yamuna-Sutlej Interfluve. Significance of terminal fan. *J Indian Soc Remote Sens* 2006, 34(2), 123–
 681 130, 10.1007/BF02991817. <u>https://link.springer.com/article/10.1007/BF02991817</u> (accessed on
 682 02/06/2017).
- Giosan, L.; Clift, P.D.; Macklin, M.G.; Fuller, D.Q.; Constantinescu, S.; Durcan, J.A.; Stevens, T.;
 Duller, G.A.T.; Tabrez, A.R.; Gangal, K.; Adhikari, R.; Alizai, A.; Filip, F.; VanLaningham, S.;
 Syvitski, J.P.M. Fluvial landscapes of the Harappan civilization. *PNAS* 2012, 109(26):
 E1688-E1694, 10.1073/pnas.1112743109. <u>http://www.pnas.org/content/109/26/E1688.full</u>
 (accessed on 02/06/2017).
- Singh, R.N.; Petrie, C.A.; Pawar, V.; Pandey, A.K.; Neogi, S.; Singh, M.; Singh, A.K.; Parikh, D.;
 Lancelotti, C. Changing patterns of settlement in the rise and fall of Harappan urbanism:
 preliminary report on the Rakhigarhi Hinterland Survey. 2009. *Man and Environment* 2010, *35*(1),
 37–53.
- 692 29. Singh, R.N.; Petrie, C.A.; Pawar, V.; Pandey, A.K.; Parikh, D. New insights into settlement along
 693 the Ghaggar and its hinterland: a preliminary report on the Ghaggar Hinterland Survey 2010.
 694 *Man and Environment* 2011, 36(2), 89–106.
- 695 30. Possehl, G.L. *Indus Age. The Beginnings*. University of Pennsylvania Press: Philadelphia, USA,
 696 1999.
- 697 31. Danino, M. *The Lost River: On the Trail of the Saraswati*. Penguin: Delhi, India, 2010.
- 32. Ramasamy, S.M.; Bakliwal, P.C.; Verma, R.P. Remote sensing and river migration in Western
 India. *Int J Remote Sens* 1991, 12(12), 2597–2609, 10.1080/01431169108955288.
 http://www.tandfonline.com/doi/abs/10.1080/01431169108955288 (accessed on 02/06/2017).
- 33. Sharma, D.C.; Kalra, N.K.; Srivastava, A.; Singh, R.B. An appraisal of space technology to
 identify the palaeochannels of river Saraswati in Anupgarh, Pilibanga and parts of Ganganagar
 and Hanumangarh districts. In *Geological Evolution of Northwestern India*; Paliwal, B.S., Ed.;
 Scientific Publishers: Jodhpur, India, 1999; pp. 260–270.

- 34. Gupta, A.K.; Sharma, J.R.; Sreenivasan, G.; Srivastava, K.S. New findings on the course of river
 Sarasvati. *J Indian Soc Remote Sens* 2004, 32(1), 1–24, 10.1007/BF03030845.
 https://link.springer.com/article/10.1007/BF03030845 (accessed on 02/06/2017).
- 35. Gupta, A.K.; Sharma, J.R.; Sreenivasan, G. Using satellite imagery to reveal the course of an extinct river below the Thar Desert in the Indo-Pak region. *Int J Remote Sens* 2011, *32(18)*, 5197–
 5216, 10.1080/01431161.2010.495093.
- 711 <u>http://www.tandfonline.com/doi/abs/10.1080/01431161.2010.495093</u> (accessed on 02/06/2017).
- 712 36. Bhadra, B.K.; Gupta, A.K.; Sharma, J.R. Saraswati Nadi in Haryana and its Linkage with the 713 Vedic Saraswati River - Integrated Study Based on Satellite Images and Ground Based 714 Information. Geol Soc India 2009, 273-288, 10.1007/s12594-009-0084-y. Ι 73, https://link.springer.com/article/10.1007/s12594-009-0084-y (accessed on 02/06/2017). 715
- 716 37. Rajani, M.B.; Rajawat, A.S. Potential of satellite based sensors for studying distribution of
 717 archaeological sites along palaeo channels: Harappan sites a case study. *J Archaeol Sci* 2011, *38*,
 718 2010–2016, j.jas.2010.08.008.
- 719 <u>http://www.sciencedirect.com/science/article/pii/S0305440310002827</u> (accessed on 02/06/2017).
- 38. Wright, R.P.; Hritz, C. Satellite remote sensing imagery: new evidence for site distributions and
 ecologies in the upper Indus. In *South Asian Archaeology 2007. Proceedings of the 19th International Conference of the European Association of South Asian Archaeology. Ravenna, Italy, 2–6 July 2007. Volume I Prehistoric Periods*; Frenez, D., Tosi, M., Eds.; Archaeopress: Oxford, UK, 2013; pp. 315–
 321.
- 39. Mehdi, S.M.; Pant, N.C.; Saini, H.S.; Mujtaba, S.A.I.; Pande, P. Identification of palaeochannel
 configuration in the Saraswati River basin in parts of Haryana and Rajasthan, India, through
 digital remote sensing and GIS. *Episodes* 2016, *39*(1), 29–38, 10.18814/epiiugs/2016/v39i1/89234.
 http://www.episodes.org/view/1339 (accessed on 02/06/2017).
- 40. van Dijk, W.M.; Densmore, A.L.; Singh, A.; Gupta, S.; Sinha, R.; Mason, P.J.; Joshi, S.K.; Nayak,
 N.; Kumar, M.; Shekhar, S.; Kumar, D.; Rai, S.P. Linking the morphology of fluvial fan systems
 to aquifer stratigraphy in the Sutlej-Yamuna plain of northwest India. *J Geophys Res-Earth* 2016,
 121, 201–222, 10.1002/2015JF003720.
- 733 <u>http://onlinelibrary.wiley.com/doi/10.1002/2015JF003720/full</u> (accessed on 02/06/2017).
- 41. Bakliwal, P.C.; Sharma, S.B. On the migration of the River Yamuna. *Geol Soc India* 1980, 21, 461–
 463. <u>http://www.geosocindia.org/index.php/jgsi/article/view/64849</u> (accessed on 02/06/2017).
- 42. Huete, A.; Didan, K.; Miura, T.; Rodriguez, E.P.; Gao, X.; Ferreira, L.G. Overview of the radiometric and biophysical performance of the MODIS vegetation indices. *Remote Sens Environ* 2002, *83*(1-2), 195–213, 10.1016/S0034-4257(02)00096-2.
 http://www.sciencedirect.com/science/article/pii/S0034425702000962 (accessed on 02/06/2017).

- 43. Gao, B. NDWI—A normalized difference water index for remote sensing of vegetation liquid
 water from space. *Remote Sens Environ* 1996, *58*(3), 257–266, 10.1016/S0034-4257(96)00067-3.
 http://www.sciencedirect.com/science/article/pii/S0034425796000673 (accessed on 02/06/2017).
- 74344. Chander, G.; Markham, B.L.; Helder, D.L. Summary of current radiometric calibration744coefficients for Landsat MSS, TM, ETM+, and EO-1 ALI sensors. *Remote Sens Environ* 2009,745113(5),893–903,10.1016/j.rse.2009.01.007.
- 746 <u>http://www.sciencedirect.com/science/article/pii/S0034425709000169</u> (accessed on 02/06/2017).
- 747 45. Zhu, Z.; Wang, S.; Woodcock, C.E. Improvement and expansion of the Fmask algorithm: cloud,
 748 cloud shadow, and snow detection for Landsats 4-7, 8, and Sentinel 2 images. *Remote Sens*749 *Environ* 2015, 159, 269–277, 10.1016/j.rse.2014.12.014.
 750 <u>http://www.sciencedirect.com/science/article/pii/S0034425714005069</u> (accessed on 02/06/2017).
- 46. Kauth, R.J.; Thomas, G.S. The tasselled cap A graphic description of the spectral-temporal development of agricultural crops as seen by LANDSAT. In *Proceedings of the Symposium on Machine Processing of Remotely Sensed Data, Purdue University of West Lafayette, Indiana, 1976;*754 Institute of Electrical and Electronics Engineers: New York, USA, 1976; pp. 4B-41–4B-51.
- 47. Crist, E.P.; Cicone, R.C. Application of the tasselled cap concept to simulated thematic mapper
 data. *Photogramm Eng Rem S* 1984, *50*, 343–352.
- 48. Crist, E.P.; Cicone, R.C. A physically-based transformation of thematic mapper data the TM tasselled cap. *IEEE T Geosci Remote* 1984, 22(3), 256–263, 10.1109/TGRS.1984.350619.
 http://ieeexplore.ieee.org/document/4157507/ (accessed on 02/06/2017).
- 49. Hudson, P.F.; Heitmuller, F.T. Local- and watershed-scale controls on the spatial variability of natural levee deposits in a large fine-grained floodplain: Lower Pánuco Basin, Mexico. *Geomorphology* 2003, 56, 255–269, 10.1016/S0169-555X(03)00155-7.
 http://www.sciencedirect.com/science/article/pii/S0169555X03001557 (accessed on 02/06/2017).
- 764 50. Cazanacli, D.; Smith, N.D. A study of morphology and texture of natural levees Cumberland
 765 Marshes, Saskatchewan, Canada. *Geomorphology* 1998, 25, 43–55,
 766 10.1016/S0169-555X(98)00032-4.
- 767 <u>http://www.sciencedirect.com/science/article/pii/S0169555X98000324</u> (accessed on 02/06/2017).
- 768 51. Brierley, G.J.; Ferguson, R.J.; Woolfe, K.J. What is a fluvial levee? *Sediment Geol* 1997, 114, 1–9, 10.1016/S0037-0738(97)00114-0.
- 770 <u>http://www.sciencedirect.com/science/article/pii/S0037073897001140</u> (accessed on 02/06/2017).
- 52. Saini, H.S.; Tandon, S.K.; Mujtaba, S.A.I.; Pant, N.C.; Khorana, R.K. Reconstruction of buried
 channel- floodplain systems of the northwestern Haryana Plains and their relation to the
 'Vedic' Saraswati. *Curr Sci India* 2009, 97(11), 1634–1643.

- 53. Dixit, Y.; Gázquez, F.; Giesche, A.; Tandon, S.K.; Saini, H.S.; Mujtaba, S.A.I.; Curtis, J.; Orengo,
 H.A.; Singh, R.N.; Hodell, D.A.; Petrie, C.A. Hydroclimate variability during the evolution of
 the Indus Civilization in northwest India. *Nat Commun* In preparation.
- 54. Neogi, S. Geoarchaeological investigations of Indus settlements in the plains of northwestern
 India. PhD thesis, University of Cambridge, Cambridge, 2013.
- 55. Neogi, S.; French, C.A.I.; Pawar, V.; Singh, R.N.; Petrie, C.A.. Geoarchaeological insights into
 the location of Indus settlements in the plains of northwest India, *Quaternary Res* in press
 (submitted 2017).
- 56. Singhvi, A.K.; Kar, A. The aeolian sedimentation record of the Thar Desert. *J Earth Syst Sci* 2004, 113(3), 371–402, 10.1007/BF02716733.
 https://link.springer.com/article/10.1007%2FBF02716733 (accessed on 02/06/2017).
- 57. Singh, G.; Wasson, R.J.; Agrawal, D.P. Vegetational and seasonal climatic changes since the last
 full glacial in the Thar Desert, northwestern India. *Rev Palaeobot Palyno* 1990, 64(1-4), 351–358,
 10.1016/0034-6667(90)90151-8.
- 788 <u>http://www.sciencedirect.com/science/article/pii/0034666790901518</u> (accessed on 02/06/2017).
- 58. Shitaoka, Y.; Maemoku, H.; Nagatomo, T. Quartz OSL dating of sand dunes in Ghaggar Basin,
 578 northwestern India. *Geochronometria* 2012, 39(3), 221–226, 10.2478/s13386-012-0012-6.
 https://www.degruyter.com/view/j/geochr.2012.39.issue-3/s13386-012-0012-6/s13386-012-0012-6
 xml?format=INT (accessed on 02/06/2017).
- 59. Maemoku, H.; Shitaoka, Y.; Natomo, T.; Yagi, H. Geomorphological Constraints on the
 Ghaggar River Regime During the Mature Harappan Period. In *Climates, Landscapes, and Civilizations*; Giosan, L., Fuller, D.Q., Nicoll, K., Flad, R.K., Clift, P.D., Eds.; American
 Geophysical Union, Geophysical Monograph 198: Washington, USA, 2012; pp. 97–106.
- 60. Durcan, J.A.; Thomas, D.S.G.; Gupta, S.; Pawar, V.; Singh, R.N.; Petrie, C.A. Holocene
 landscape dynamics in the Ghaggar-Hakra palaeochannel region at the northern edge of the
 Thar Desert, northwest India, *Quatern Int* In press (submitted 2016).
- 800 61. Conesa, F.C.; Devanthéry, N.; Balbo, A.L.; Madella, M.; Moserrat, O. Use of satellite SAR for 801 understanding long-term human occupation dynamics in the monsoonal semi-arid plains of 802 North Gujarat, India. Remote Sens. 2014, 11420-11443, 10.3390/rs61111420. 6, 803 http://www.mdpi.com/2072-4292/6/11/11420 (accessed on 02/06/2017).
- 62. Langford-Smith, T. Riverine Plains geochronology. *Australian Journal of Science* **1962**, *25*, 96–97.
- 805 63. Singh, R.N.; Petrie, C.A.; French, C.A.I.; Goudie, A.S.; Gupta, S.; Tewari, R.; Singh, A.K.; Sihna,
 806 R.; Srivastava, R.; Yadav, S.; Singh, V.K. Settlements in Context: Reconnaissance in western
 807 Uttar Pradesh and Haryana. *Man and Environment* 2008, 33(2), 71–87.

- 64. Green, A.S.; Petrie, C.A. Landscapes of urbanisation and de-urbanization: integrating site
 location datasets from northwest India to investigate changes in the Indus Civilization's
 settlement distribution. In prep. (to be submitted 2017).
- 65. Clift, P.D.; Carter, A.; Giosan, L.; Durcan, J.; Duller, G.A.T.; Macklin, M.G.; Alizai, A.; Tabrez,
 A.R.; Danish,M.; VanLaningham, S.; Fuller, D.Q. U-Pb zircon dating evidence for a Pleistocene
 Sarasvati River and Capture of the Yamuna River. *Geology* 2012, 40, 211–214, 10.1130/G32840.1.
 http://geology.gsapubs.org/content/early/2012/01/23/G32840.1.abstract (accessed on
 02/06/2017).
- 816 66. Google Earth Engine Team. 2015. Google Earth Engine: A planetary-scale geo-spatial analysis
 817 platform. <u>https://earthengine.google.com</u> (accessed on 02/06/2017).
- 818 67. Agapiou, A. Remote sensing heritage in a petabyte-scale: satellite data and heritage Earth
- 819 Engine© applications. Int J Digit Earth 2017, 10(1), 85–102, 10.1080/17538947.2016.1250829.
- 820 <u>http://www.tandfonline.com/doi/full/10.1080/17538947.2016.1250829</u> (accessed on 02/06/2017).



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