1 2	Evolutionary, multi-scale analysis of river bank line retreat using continuous wavelet transforms: Jamuna River, Bangladesh.
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20 Abstract

In this study continuous wavelet transforms are used to explore spatio-temporal 21 22 patterns of multi-scale bank line retreat along a 204 km reach of the Jamuna River, 23 Bangladesh. A sequence of eight bank line retreat series, derived from remotely-sensed 24 imagery for the period 1987-1999, is transformed using the Morlet mother wavelet. 25 Bank erosion is shown to operate at several characteristic spatial and temporal scales. 26 Local erosion and bank line retreat are shown to occur in short, well defined reaches 27 characterised by temporal persistence at the same location, and separated by relatively 28 stable reaches. In contrast, evidence of downstream propagation of bank line retreat 29 patterns is evident at larger spatial scales. The intensity of localised bank line retreat 30 (i.e. at scales of 0 - 20 km) is strongly related to the magnitude of monsoonal peak 31 discharge, but this relationship weakens as the spatial scale of erosion increases. The 32 potential of continuous wavelet analysis to enhancing our understanding of 33 morphological evolution in complex fluvial systems with multi-channel planforms is 34 discussed.

35 Keywords

Continuous Wavelet Transform, Jamuna River, Braided river, Time-space, Erosion
 processes, Embayment pattern, Sediment wave.

38 **1. Introduction**

39 The planform evolution of the Jamuna River, Bangladesh, the distal portion of the 40 Brahmaputra, has been the focus of many studies in which the aim has been to assess, 41 characterise and quantify the magnitude and distribution of bank line migration (e.g. 42 Coleman, 1969; Sarma and Basumallick, 1984; Singh et al., 1990; Thorne et al., 1993, 43 1995; Thorne and Russell, 1993; Halcrow, 1994; Goswami, 1995; Goswami et al., 1999; 44 Ashworth et al., 2000; CEGIS, 2000, 2007; Khan and Islam, 2003; Sarma and Phukan, 45 2004, 2006; Sankhua et al., 2005; Sarma, 2005; Takagi et al., 2007). However, 46 quantification and prediction of the spatio-temporal patterns of bank line migration

47 exhibited by the river are complicated by its large geographical scale (here synonymous 48 with extent) and its complex planform, which features elements of meandering, braiding 49 and anastomosing (Fergusson, 1993). The large geographical scale has thus far 50 precluded the application of well-established hydraulic geometry relationships or 51 process-based explanatory models due to the difficulties associated with up-scaling 52 (Latrubesse, 2008). Furthermore, the river's complex and dynamic planform has also 53 precluded application of conventional, geometric models developed for simpler, 54 meandering channels (e.g. Ikeda et al., 1981; Parker et al., 1983; Johannesson and 55 Parker, 1989; Zolezzi and Seminara, 2001; Camporeale et al., 2005). In fact, the 56 difficulty inherent in large-scale, long-term studies of channel evolution and bank line migration in complex, multi-channel rivers prior to the wide availability of high definition, 57 58 remotely sensed imagery explains why such studies have, until recently, been rarely 59 conducted (Best and Bristow, 1993; Richardson, 1997). To date, the majority of 60 geomorphological studies of the Jamuna have instead focused on investigating the processes responsible for channel evolution and bank line migration at the scale of the 61 62 individual geomorphological unit. Examples include studies of channel bifurcations and 63 braid bars (Ashworth et al., 2000; Richardson and Thorne, 2001); braid bars and 64 associated floodplain embayments (Thorne et al., 1993; Halcrow, 1994) and the 65 evolution of meander bends in major anabranches (Thorne and Russell, 1993; Ellis, 66 1993).

67 The bank line adjustment processes associated with these different geomorphological units exhibit non-stationarity (spatial and temporal localization) as 68 69 well as different, characteristic bank migration rates and scale-dependency in space (circa $10^2 - 10^4$ m) and time (circa $10^0 - 10^1$ yr). It follows that, in rivers with multi-70 71 channel planforms, the overall pattern of bank retreat is characterized as a complex non-72 stationary waveform within which multiple, characteristic erosion patterns co-exist at 73 different scales, and at different downstream locations. For example, Thorne et al. 74 (1993) identified a gross-scale control of planform evolution associated with island and

75 nodal reaches first described by Coleman (1969) that are spaced along the channel at 76 intervals of circa 30 km. They also identified local bank migration processes that are 77 driven by braid bar growth and migration and which operate at smaller spatial (3 - 6 km) and over shorter temporal scales (2 - 5 years). Their findings contrast with those of Ellis 78 79 (1993), who observed bank erosion and embayment formation related to meander 80 growth and migration in near-bank anabranches that occur at spatial scales of hundreds 81 of metres to several kilometres, persist over periods of 1-12 years, and drive erosion rates ranging from 50 to over 250 m yr⁻¹. More recently, CEGIS (2007) identified bank 82 83 erosion and floodplain embayment forming processes associated with bar form 84 development in the Jamuna river operating at spatial scales of 3-15 km, over periods of 85 approximately 15 years and with erosion rates of of the order of 200 m yr⁻¹. A finding 86 common to these studies was that rates of downstream migration in the locations of 87 severe bank erosion also appear to be scale-dependent. This is consistent with the 88 downstream movement of sand bars (Coleman, 1969) and changes in the location of 89 relatively stable and unstable reaches over time (Takagi et al., 2007), which all suggest 90 a link to the downstream propagation of sediment waves (Gilbert, 1917; Madej and 91 Ozaki, 1996; Wathen and Hoey, 1998). Indeed both Thorne et al. (1993) and Takagi et 92 al. (2007) identified wave-like patterns of channel migration, with characteristic 93 wavelengths of ~150 km and ~35 km, respectively. These findings suggest that the 94 complex patterns of bank retreat observed in the Jamuna result from the superimposed 95 and cumulative effects of spatially-transient bank erosion processes, operating semi-96 independently at a range of spatial and temporal scales.

97 Marcus and Fonstad (2010) argue that the development and availability of new 98 remote sensing technologies, coupled with widening accessibility to GIS, have led to the 99 emergence of the 'remote sensing of rivers' as a sub-discipline of fluvial geomorphology. 100 Complex patterns of bank migration are now commonly investigated based on temporal 101 sequences of bank line data, captured from aerial photographs or remotely-sensed 102 imagery. These data are analysed within a geographical information system (GIS) so

103 that rates of bank line retreat during specified periods can be computed (e.g. Gurnell et 104 al., 1994; Mount et al., 2003; Mount and Louis, 2005; Swanson et al., 2011). In such 105 studies, characterisation of downstream migration in bank erosion patterns reduces to a 106 problem of localizing in space the different magnitudes and scales of bank retreat events 107 that can be discerned from the data, and evaluating how these localized patterns vary 108 through time. In the past this has been achieved through a largely qualitative, visual 109 appraisal of the patterns observed in sequential plots of bank position or change (e.g. 110 Downward et al., 1994; Gurnell, 1997; Mount et al., 2003) coupled with examination of 111 summary bank retreat statistics (e.g. CEGIS 2000; 2007). Consequently, relatively little 112 quantitative understanding of either trends in the spatial localization associated with 113 erosion at different scales or the relative rates of erosion driven by the different 114 processes represented in the bank retreat record has been achieved. Moreover, it has 115 been difficult to causally relate the observed patterns of bank retreat to likely 116 geomorphological drivers; especially where these involve processes acting at different 117 spatial and temporal scales. Commenting on this in a recent review, Kleinhans (2010) 118 acknowledged the potential of remotely sensed time-series for unravelling channel 119 pattern changes, but also observed that, 'we need quantifiers for subtle patterns to 120 reveal structure objectively' (Kleinhans, 2010; page 313).

121 A more quantitative approach involves viewing the downstream distribution of 122 bank retreat as a spatial signal of planimetric change. According to this approach, the 123 different scales of erosion are equivalent to the different frequencies contained within the 124 signal and their magnitudes are equivalent to signal amplitude. Accordingly, the 125 downstream signal of erosion becomes the spatial equivalent of a standard time series 126 signal, with distance substituted for time. When re-conceptualised in this way, 127 techniques developed for the characterisation of frequency and amplitude in complex 128 signals, become potentially powerful tools for characterising bank retreat. One common 129 tool is the fast Fourier Transform (FFT), which offers good frequency localisation albeit at the expense of poor spatial localisation (Graps, 1995). However, the FFT requires the 130

131 data series to be consistent with a statistically stationary model – something that cannot 132 necessarily be assumed when analysing bank retreat sequences in river channels (Van 133 Gerven and Hoitink, 2009). Alternative methods offering both frequency and spatial 134 localization, such as the windowed Fourier transform (WFT) (also known as the short-135 time Fourier transform) and wavelet analysis are therefore, preferable. Both have been 136 applied to the analysis of river patterns (Ferguson, 1975; Camporeale et al., 2005, Van 137 Gerven and Hoitink, 2009). However, the WFT is considerably less adaptable than 138 wavelet analysis in that achieving good spatial resolution requires sacrificing frequency 139 localisation, and vice versa (Fournier, 1995 page 11). As a result WFT was not selected 140 for use in this research. While the use of wavelet analysis by geomorphologists remains 141 rare, it has been widely used by hydrologists for the characterisation of runoff time 142 series (e.g. Brillinger, 1994; Fraedrich et al., 1997; Labat et al., 1999, 2000, 2002; 143 Compagnucci et al., 2000; Gaucherel, 2002; Lafreniere and Sharp, 2003; Coulibaly and 144 Burn, 2004; Labat, 2005) Also, its potential in analysing signals in time series of river 145 planform change has recently been recognised by Van Gerven and Hoitink (2009), who 146 used it to characterise meander geometry on the Mahakam River, Indonesia. There are, 147 however, no published examples of its use in characterising bank migration in large 148 rivers with anabranching planforms.

149 In this paper we explore the potential of the continuous wavelet transform (CWT) 150 for the quantitative characterisation of temporal sequences of downstream bank line 151 migration patterns, using data recorded along a 204 km reach of the Jamuna River, 152 Bangladesh. The CWTs presented in this paper quantify changes in bank position in the 153 plane of maximum bank erosion. In line with previous studies of planform change from 154 remotely-sensed data, this is assumed to be orthogonal to the downstream direction. 155 This is believed to be the first time such an approach has been applied in a bank line 156 retreat study. In section 2 we describe the CWT method. Section 3 presents the 157 geomorphologic application, results and interpretation of the CWT to patterns of bank

line retreat on the Jamuna River. The key findings from the study are summarised insection 4.

160 2. Methodology

161 Wavelet analysis comprises several mathematical transforms from which 162 temporal (or spatial) series can be transformed into a 2-D time (or space)-frequency 163 representation. A detailed treatment of the mathematics of wavelet transforms is 164 beyond the scope of this paper, but may be found in publications covering both wavelet 165 analysis theory (see Daubechies, 1992) and software implementation (Nason, 2008). 166 In this paper we make use of the continuous wavelet transform (CWT) popularised by 167 Torrence and Campo (1998). Our description of the CWT below draws from Torrence and Campo (1998) using the notation and development of Sadowsky (1996), Kumar and 168 169 Fourfoula-Georgiou (1997), and Biswas and Si (2011). In this description, y(x) denotes 170 the spatial series of left bank (LB) downstream distance measurements. We can define the CWT, which we denote by $W(s, \sigma)$, as the complex conjugation of y(x) with a dilated 171 172 and translated `mother' wavelet function $\psi_{s,\sigma}(x)$:

$$W(s,\sigma) = \int_{-\infty}^{\infty} y(x) \overline{\psi_{s,\sigma}}(x) d(x)$$

(1)

(2)

173

174 where

 $\psi_{s,\sigma}(x) = \frac{1}{\sqrt{s}}\psi\left(\frac{x-\sigma}{s}\right)$

175

176 Here, *s* represents the dilation (scale) of the wavelet function and σ represents the 177 degree of distance translation along the series. The term $1/\sqrt{s}$ normalises the wavelet 178 function energy at each scale (Kumar and Foufoula-Georgiou, 1997; Torrence and 179 Campo, 1998). Equation (2) emphasises that the wavelet function is in fact a 'basis' 180 function (Fournier, 1995). For the CWT of a discretely sampled spatial series denoted 181 by $W_D(s)$ for a distance index *d* the integral in Equation (1) is substituted by a 182 summation and the distance *x* is replaced by increments of size δx (Torrence and Campo, 183 1998; Gurley and Kareem, 1999; Biswas and Si, 2011). Finally, the wavelet power 184 spectrum for a given transform can be defined as $|W_D(s)|^2$ (Torrence and Campo, 1998; 185 Biswas and Si, 2011).

186 Wavelet functions are required to have a compact support (in other words they 187 decrease rapidly to zero) and a mean of zero (Farge, 1992; Kumar and Foufoula-188 Georgiou, 1997). In spite of these requirements, there are a great number of functions 189 available which satisfy these criteria. These can be classified in various ways such as (a) 190 orthogonal or non-orthogonal; (b) complex or real and in terms of (c) width and (d) 191 shape (see Torrence and Campo, 1998 § 3e). *Non-orthogonal* wavelets (they overlap) 192 are used in the CWT. The CWT therefore incorporates considerable redundancy in the 193 representation of the spatial series (Kumar and Foufoula-Georgiou, 1997) and an 194 alternative to this is to make use of orthogonal wavelets which are the basis of the 195 discrete wavelet transform (DWT) which in its simplest form can be thought as a set of 196 slices through the CWT at scales defined by powers of two (Percival et al., 2004). 197 However, it is argued that the non-orthogonal wavelets as used in the CWT are possibly 198 more appropriate for spatial series analysis as they can reveal more information on scale 199 localization (Biswas and Si, 2011). Non-orthogonal versions of the DWT do exist such as 200 the maximal overlap discrete wavelet transform and maximal overlap discrete wavelet 201 packet transform which have also been applied in the analysis of spatial series (Milne et 202 al., 2010).

203 Common examples of complex non-orthogonal wavelets used in the CWT (and 204 the focus of Torrence and Campo, 1998) include the Morlet (Fig. 1 A,B, Equation 3) and 205 the Paul (Fig. 1 C, Equation 4) wavelets, whilst the Derivative of Gaussian (DOG) (Fig. 1 D, Equation 5) is an example of a real-valued function (equations modified fromTorrence and Campo, 1998):

$$\psi(x) = \pi^{-1/4} e^{ikx} e^{-x^2/2}$$
(3)

(4)

208

209

 $\psi(x) = \frac{(-1)^{k+1}}{\sqrt{\Gamma\left(k+\frac{1}{2}\right)}} \frac{d^k}{dx^k} \left(e^{-x^2/2}\right)$

 $\psi(x) = \frac{2^k i^k k!}{\sqrt{\pi(2k)!}} (1 - ix)^{-(k+1)}$

210	(5)
10	(5)

Here, *k* is the parameter (known as the wavelet order) that controls the number of oscillations in the wavelet function with the result of altering the resolution (both frequency and distance) of the wavelet transform (De Moortel et al., 2004).

214 The choice of the most appropriate mother wavelet function to use in a wavelet 215 analysis (along with the optimal value of k) depends largely on the characteristics of the 216 signal itself, ideally reflecting the shapes of features present in the data series to be 217 analysed (Lane, 2007). However, there is a distinct lack of advice in the wavelet 218 literature on the optimal choice of these parameters, leading to the development of 219 context dependent criteria for relative comparison (e.g., Fu et al., 2003) which may or 220 may not be universal. Although Torrence and Campo (1998) have suggested that 221 different functions will nevertheless give the same qualitative results for wavelet power 222 spectra (Torrence and Campo, 1998), the practical advice of De Moortel et al. (2004) to 223 experiment with different parameters would seem appropriate.

224 2.1. Cone of influence and significance levels

As noted by De Moortel et al., (2004) the CWT as implemented in code such as 225 226 that developed by Torrence and Campo (1998) is often speeded up by transforming to 227 Fourier space. Common to Fourier transforms of finite series this introduces edge effects 228 primarily due to 'spectral leakage' as a result of edge discontinuities (Fougere, 1985). 229 There are a variety of methods to ameliorate such effects: the approach taken in 230 Torrence and Campo (1998) is 'zero padding' whereby zeros are added to the data series 231 up to the next integer power of two. This results in a 'cone of influence' (COI) at the 232 margins of the wavelet transform where the interpretation of the wavelet transform 233 should be considered unreliable (De Moortel et al., 2004).

234 It is also possible to perform a statistical significance test and calculate 235 significance by comparing the wavelet power to an appropriate background noise 236 spectrum. Commonly, either white noise or red noise (increasing power with decreasing 237 frequency) are used with the latter being a more realistic model of many geophysical 238 series that exhibit short distance spatial dependence (see Fougere, 1985). Red noise is 239 therefore often preferred (e.g., Si and Farrell, 2004 and references therein) and can be 240 modelled by a first order autoregressive AR(1) process (modified from Torrence and 241 Campo, 1998):

$$x(t) = c + \alpha x(t-1) + z(t)$$

(6)

242

243 Where *c* is a constant, α is the lag-1 autocorrelation and z(t) is white noise. We refer 244 the interested reader to the description of the Fourier power spectrum of (6) and a full 245 explanation with formulae for the quantification of significance levels using a Monte Carlo 246 simulation approach to Torrence and Campo (1998; § 4) as well as a summary 247 treatments by Si and Farrell (2004). If a peak in the wavelet power spectrum is 248 significantly above this background we might ascribe this to a scale dependent 249 pattern/process operating at that frequency. In this study we use a red noise 250 background spectrum with α computed according to the AR(1) coefficient and present all 251 significant results using the 95% confidence level.

252 2.2. Scale-averaging wavelet spectra

The fluctuations in wavelet power across discrete scale ranges or bands can be achieved by defining the *scale-averaged wavelet power* as the weighted sum of the wavelet power spectrum over scales s_1 to s_2 (modified from Torrence and Campo, 1998; Coulibaly and Burn, 2004):

$$\overline{W}_D^2 = \frac{\delta j \delta x}{C_\delta} \sum_{j=j_1}^{j_2} \frac{\left|W_D(s_j)\right|^2}{s_j}$$

(7)

257

where C_{δ} is a constant that can be derived for any wavelet function via reconstruction, and δj depends on the width of the wavelet function used and should ensure adequate sampling in scale (see Torrence and Campo, 1998 § 5b for the derivation formulae). Significance levels can also be ascribed to the scale averaged wavelet power via an analytical relationship between the significance levels and the scale-averaged wavelet power. For details of this the reader is again directed to Torrence and Campo (1998).

264 The scale averaged wavelet power is a series of the average variance in a certain band.

265 It can, therefore, be used to examine modulation of one series by another and / or

266 modulation of one frequency series by another within the same series.

3. Geomorphologic application: bankline retreat characterisation of the Jamuna River

In this study we analyse bank migration along a 204 km reach of the Jamuna river in Bangladesh (Fig. 2), between its confluence with the Teesta River just south of the Indian border, and the Ganges River. The Jamuna is one of the largest and most dynamic rivers in the world ranking fifth in terms of discharge (mean flow 12,200

273 cumecs) and eleventh in terms of drainage area ($666,000 \text{ km}^2$) (Thorne et al., 1993). 274 Analysis of the long-term evolution of the channel (Coleman, 1969; Burger et al., 1991; 275 Thorne et al., 1993; ISPAN, 1995; CEGIS, 2001; Takagi et al., 2007) has revealed a 276 highly dynamic channel that has undergone westward migration, widening and planform 277 metamorphosis following the creation of the present river by avulsion in 1830. The 278 meandering planform of the 19th century channel has been replaced by a much wider, 279 braided channel throughout the 20th century, and this change has been accompanied by 280 very high rates of bank line retreat. In the last three decades, the majority of this 281 retreat can be related to channel widening rather than centreline migration, although the 282 left hand bank does show a slightly higher average retreat rate (68 m yr⁻¹ between 1973 and 2000) than the right hand bank (60 m yr⁻¹) (CEGIS, 2000). 283

284 In this study we concentrate our analysis on patterns of bank retreat occurring 285 between 1987 and 1999. This period is characterised by particularly rapid channel 286 widening with spatially-variable and non-stationary patterns of stabilisation / 287 destabilisation; possibly associated with propagating sediment waves (Thorne et al., 1993; Takagi, 2007), related to sediment supplied from upstream by the 1950 288 289 earthquake in Assam (Goswami et al., 1999). The pattern of bank retreat evidences 290 erosion at a wide range of scales from individual embayments to island reach scales. The 291 existence of localised, multi-scale bank retreat patterns, coupled with high retreat rates 292 make the period particularly well suited to wavelet analysis.

293 The floodplain of the left-hand bank (LHB) is comprised of newly accreted land 294 which is characterised by poorly consolidated and easily eroded bank material. This has 295 resulted in much more localised and complex patterns of bank retreat than that of the 296 right-hand bank (RHB) (Thorne et al., 1993). Indeed, between 1973 and 2006, 49,460 297 hectares of land eroded along 190 km of LHB, compared to 38,540 hectares along 245 298 km of RHB (CEGIS, 2007). Similarly, maximum rates of erosion are highest on the LHB with local erosion exceeding 2000 m yr⁻¹ in several locations. As a consequence, the 299 300 downstream pattern of bank retreat for the LHB is characterised by events of greater

amplitude, greater variability of scale and greater spatial variability than the RHB, and
we have therefore selected the LHB data for subsequent analysis in this study.

303 *3.1. Bank line delineation methods and data series*

304 The data used in this study have been sourced from the Centre for Environmental 305 and Geographical Information Services (CEGIS), Dhaka. CEGIS have been responsible 306 for quantifying bank retreat rates along the entire length of the Jamuna in Bangladesh 307 from satellite imagery for the period 1973 - present. Their data has formed the baseline 308 geomorphological dataset for the World Bank's Flood Action Plans (FAP-1, 1991; 1992). 309 We provide a summary below of the methods used to generate the data – further details 310 can be obtained from CEGIS (1997, 2001), including information on ground validation of 311 satellite image analysis.

312 A time-series of dry season Landsat MSS and TM images were used to document 313 historical changes in LHB position along the study reach. The time series of images 314 covered the period 1987-1999 (Table 1), allowing LHB migration to be computed for 8 315 separate periods: 1987-89; 1989-92; 1992-94; 1994-95; 1995-96; 1996-97; 1997-98 316 and 1998-99. All images were mosaiced and georeferenced to a 1:50,000 scale colour base map; originally derived from 1989 high resolution SPOT satellite images, and 317 318 projected using the UTM-46N projection (CEGIS, 2007). For each image in the mosaic, 319 more than 25 ground control points were used to define a first order transformation to 320 the base map. Where possible, ground control points were taken from recognizable and 321 permanent features such as road intersections, airport runways and large buildings. The 322 maximum root mean square (RMS) error of the transformation was 96 m for the Landsat 323 MSS imagery (1987) and 45 m for the Landsat TM imagery (1989-1999).

The Jamuna in Bangladesh flows approximately due south (Fig. 2), and the channel is therefore orientated along a consistent northing ordinate throughout its length. Following the established methods of Gurnell et al. (1994); Gurnell (1997); Mount et al. (2003) and Mount and Louis (2005) maximum bank erosion in the Jamuna can therefore

328 be assumed to occur in a direction orthogonal to the main channel (i.e. in an easterly 329 direction). Bank line position was assessed at a downstream spacing of 500 m; defined 330 by each image's northing coordinate. Consequently, LHB migration between consecutive 331 images at each downstream location was computed as the difference in the easting coordinate of the bank line position, converted to m yr⁻¹ according to the capture dates 332 333 of the imagery (Table 1). Bank lines were defined as the line which separates the 334 floodplain from the active braid belt. In general, bank lines encompass main channels, 335 island chars and sand bars in the river braid belt, except for crevasse splays (the coarse 336 sediments that are spread over the floodplains during floods). Where a major anabranch 337 of the river flows along the edge of the floodplain in a channel (which typically ranges in 338 width from hundreds of meters to several kilometres), the bank line delineation is simple 339 and uncontroversial. However, smaller distributary channels flowing next to the bank are 340 more difficult to define and a number of criteria can be applied to determine whether 341 they should be considered a part of the active braid belt (CEGIS, 1997, 2000):

342 (1) channels are outside the bank line if the channel does not return to the main river;

343 (2) channels are outside the bank line if the channel is less than 100 m in width;

(3) channels are outside the bank line if the channel has a meander radius of less thanone kilometre.

346 Due to the difficulties associated with determining whether material deposited at 347 the channel margins between images remains active channel or has become 348 incorporated into the floodplain, only bank line retreat rates are recorded in the data 349 series (deposition is recorded as zero). This means that the analysis presented here 350 represents a partial examination of the patterns on planimetric change on the Jamuna, 351 which comprise both bank line retreat and bank advance; the latter occurring through 352 the incorporation of sediments previously considered to be within the active channel into 353 the floodplain. All bank line retreat between consecutive images of 50 m or less was 354 considered to be within the margin of measurement and georeferencing error and was

355 recorded as zero retreat. For the periods 1989-92, 1992-94; 1994-95; 1995-96; 1996-356 97; 1997-98 and 1998-99, where all images are of 30 m resolution, this margin exceeds 357 the 42 m quadratic sum associated with a ground feature identification error of ± 1 pixel 358 in each image (c.f. Mount and Louis, 2005). For the period 1987-89, in which one 30 m 359 and one 80 m resolution image are used, this value is less than the 85 m quadratic sum 360 of the ± 1 pixel ground feature identification error. Consequently, the magnitude of bank line retreat in this period may be slightly over-estimated in the data. In addition to 361 the bank line data, daily mean discharge records from Bahadurabad (21.15°N, 89.70°E) 362 were also acquired from CEGIS. The resultant temporal sequence of bank line retreat 363 364 series is provided in Figure 3.

365 *3.2.* Wavelet function selection

366 In this study we consider the three wavelet functions described in Torrence and 367 Campo (1998) and detailed in Equations (3-5). These three functions offer a range of 368 different localisation capabilities (which are described in detail in De Moortel et al. 2004). 369 The Morlet wavelet (Equation 2) offers high frequency localisation capabilities which 370 result from the presence of a large number of oscillations but this increases the wavelet 371 width which in turn reduces spatial localisation capability. The Paul (Equation 3) wavelet 372 offers higher spatial resolution (it has fewer oscillations), but at the expense of 373 frequency localisation. The DOG (Equation 4) offers excellent spatial localisation of 374 individual peaks in the series, but can suffer from discontinuity in the frequency 375 localisation in the transform.

To assist in the identification of the preferred wavelet, a CWT decomposition of the 1998-99 bank line retreat data was undertaken using all three wavelet functions using the software developed by Torrence and Campo (1998) and implemented in MATLAB. The 1998-99 data was selected because it represents one of the more complex data series in the temporal sequence with a wide range of spatial scale and magnitudes of bank line retreat evident. The relative lack of local oscillation complexity in the bank 382 line retreat series implies adequate frequency localisation should be achieved with a 383 relatively low wavelet parameter (k) value. Moreover, the use of a small value of k384 reduces the area of the transform under the COI. Therefore, k was set to 3. However, a 385 transform was also generated using a Morlet, k = 6 wavelet to determine the impact of 386 increasing the k value. The results are provided in Figure 4.

387 The Paul k=3 and Morlet k=3 wavelets can be seen to generate very similar patterns of significant wavelet power, associated with the same local peaks of bank line 388 389 retreat. There are subtle differences between the two in terms of their space / 390 frequency localisation capabilities. The Morlet k=3 transform exhibits some horizontal 391 compression and increased lateral connectivity of the 95% confidence regions in 392 comparison with the Paul k=3 transform; thereby highlighting its greater frequency 393 localisation capabilities. The DOG wavelet can be seen to localise peaks particularly well, 394 but at the expense of frequency. The result is lateral discontinuity in the 95% 395 confidence regions (which is also reported in De Moortel et al., 2004) and a frequency 396 localisation that is difficult to interpret. Increasing the Morlet wavelet k value to 6 397 results in a larger COI and a loss of significant wavelet power regions at high frequencies 398 and the loss of spatial localisation (i.e. the 95% confidence regions elongate 399 horizontally); thus making it difficult to map the significant bank line retreat frequencies 400 accurately on the ground. The results indicate that both the Paul k = 3 and Morlet k = 3401 represent appropriate wavelet functions for the bank line retreat data series. However, 402 in this paper we use the Morlet k = 3 wavelet on the basis that it provides enhanced 403 frequency localisation with minimal loss of spatial localisation.

404 *3.3. Results*

405 3.3.1. Continuous wavelet transform spectra

406 The CWT spectra for the 8 bank line retreat series (Fig. 3) calculated using an 407 adaptation of the Torrence and Campo (1998) code are presented in Figure 5. Regions 408 of wavelet power that exceed the 95% confidence level are those within the bold lines,409 with the dashed line indicating the COI.

410 The CWT spectra highlight two characteristic groupings of bank line retreat 411 patterns over the study period. Zones of significant wavelet power in 1996-97; 1997-98 412 and 1998-99 are largely constrained to wave periods of 16 km or less. These comprise a 413 mixture of small, very short wave periods (1-4 km) with very high spatial and frequency 414 localisation and a broader range of wave periods (2-16 km) in which the spatial and frequency localisation is less well resolved. At both wave periods, significant zones are 415 416 spatially discrete and separated by downstream distances of between 10 and 20 km. 417 Moreover, there appears to be some temporal persistence in their location. Such 418 patterns are characteristic of locally-persistent bank line retreat events which occur 419 independently of any larger bank line retreat process (i.e. there is little evidence of local 420 bank line erosion patterns being superimposed on larger-scale, regional patterns). In 421 certain cases, the retreat is highly constrained to within a very short stretch of bank (i.e. 422 the 1-4 km wave period regions), whereas in others longer stretches are implicated (2-423 16 km period regions).

424 By contrast, zones of significant wavelet power in 1987-89; 1989-92; 1992-94 425 1994-95 and 1995-96 are characterised by the existence of longer wave periods (16-64 426 km) extending over substantial downstream distances. These are coupled, to varying 427 extents, with superimposed and locally-discrete short wave period zones, which are 428 similar in character to those of 1996-97; 1997-98 and 1998-99, but less numerous. 429 Importantly, there is some evidence of downstream movement in the locations of the 430 longer wave period significance zones, from 0 – 60 km in 1987-89; 40 – 80 km in 1992-431 94; 60 – 100 km in 1995 to 95 and 100 – 160 km in 1995-96. Such patterns are 432 difficult to interpret, yet their characteristic long wave periods, coupled with substantial 433 downstream extension and translation, would indicate that they may be characteristic of 434 bank retreat driven by the passage of a sediment-wave, or linked to the presence of 435 large, island char (quasi-stable, mature, vegetated islands within the active channel belt).

436 *3.3.2.* Scale averaging

437 Each of the CWTs was scale averaged into 0-10 km and 10-30 km bands 438 according to the scales of the two major downstream controls of bank retreat pattern 439 identified as operating on the Jamuna (c.f. Thorne et al., 1993). The 0-10 km band 440 encompasses the scale of individual braid-bars responsible for local bank retreat due to 441 embayment (CEGIS, 2007) and local bend evolution (Ellis, 1993). The 10-30 km band 442 encompasses the scale of island / nodal reaches associated with lower frequency 443 patterns of bank retreat governing gross-scale planform evolution (Thorne et al., 1993), 444 and the wavelength of propagating sediment waves (Takagi et al., 2007). The 445 downstream pattern of scale-averaged wavelet power exceeding the 95% confidence 446 level for each time period is plotted onto a single graph for each scale range (Fig. 6). 447 This allows the spatial and temporal persistence of different scales of bank retreat to be 448 investigated. Because the 95% confidence level of the scale-averaged wavelet power 449 varies between each time period, y-axis values are standardised as multiples of the 95% 450 confidence level for each retreat period.

451 At the 0-10 km scale, the LHB can be separated into a number of clearly defined 452 reaches according to the magnitude (i.e. the wavelet power multiples above the 95% 453 confidence level) and spatio-temporal persistence (i.e. the number of consecutive years 454 wavelet power exceeds the 95% confidence level at a given location) of the bank retreat 455 patterns (Table 2). Three low-magnitude, stable reaches, which exhibit little or no 456 significant wavelet power at scales of 0-10 km, at any time period, are visible at ~57-65 457 km, ~120-135 km and ~155-164 km. The last of these is almost certainly related to the 458 existence of stable guide bunds for Jamuna Bridge (Fig. 2). In all cases, these stable 459 reaches are short; not exceeding 15 km in length. In contrast two reaches exhibit 460 persistent significant 0-10 km wavelet power throughout the majority of the study period, which is, at times, of high magnitude. These are located at ~65-85 km and ~135-155 461 462 km. They represent reaches of moderate length (~20 km), which exhibit consistent and 463 substantive bank line migration operating at scales indicative of embayment and

464 meander bend processes. Importantly, there is little evidence of downstream translation 465 of bank line retreat with the locations of each of the main peaks in wavelet power 466 occurring within ~ 10 km of each other (i.e. within the bounds of the spatial localisation 467 capabilities of the Morlet k=3 mother wavelet). Between these two end members are a 468 further three reaches in which the spatio-temporal pattern of bank retreat is transient 469 and of variable magnitude. These reaches range in length from 18 - 39 km. 470 Importantly, the transient reaches show clear evidence of downstream translation of the 471 peaks in wavelet power, as indicated by arrows in Figure 6. This may imply the 472 presence of transfer reaches where sediment eroded from the persistent, high 473 magnitude reaches is transferred downstream; instigating localised bank retreat as it is 474 transported.

475 At the island / node reach scale (10-30 km) there is little temporal persistence in 476 the locations of peak wavelet power; and thus of bank line retreat. However, the 477 existence of a stable, nodal reach at 100 - 120 km is of note. There is clear evidence of 478 downstream propagation of bank retreat. Between 1987-89 and 1995-96 a consistent 479 downstream translation of the peak wavelet power is visible, from ~25 km to ~75 km, 480 and at a mean annual rate of approximately 8 m yr⁻¹. This pattern is consistent with 481 that of sediment wave propagation observed by Takagi et al. (2007), both in terms of its 482 scale and location.

483 The temporal analysis presented in Figure 6 provides a simplified appraisal of the 484 pattern of location and frequency variation in bank retreat through time, where the 485 complexity has been reduced through the temporal isolation of the scale-averaged power 486 spectra from each CWT. As a result, the analysis does not provide a detailed evaluation 487 of the patterns of covariance from one CWT to the next. Techniques such as cross-488 wavelet transforms and wavelet coherence analysis (Grinsted et al., 2004), that 489 specifically focus on quantifying the covariance between sequences of CWTs, offer 490 considerable potential in this regard (Torrence and Webster, 1999). Whilst the

491 application of these advanced techniques is beyond the scope of this paper, their492 importance as a direction for future research should be recognised.

493 *3.3.3.* Wavelet power relationship to discharge

One of the key challenges in geomorphologic studies has been establishing and elucidating the relationships between processes operating over different temporal and spatial scales (Rhoads and Thorn, 1996, pp. 145-6; Phillips, 1999a,b; Couper, 2004). Wavelet decomposition of sequential geomorphologic signals offers potential in this regard through the examination of the pattern of variability in the strength of the wavelet power at different frequency localisations and time periods and its relationship with potential physical drivers of the signal.

501 To this end, we also examine the relationship between different scales of bank 502 retreat and discharge by quantifying the definite integral of the downstream scale-503 averaged wavelet power spectra each time period (i.e. a numerical proxy for the total 504 magnitude of LHB retreat occurring at that scale) and plotting this against two measures of peak discharge: the maximum discharge and the Q^{95} exceedance period (Table 3). 505 Whilst maximum discharge is identified as an important driver of bank retreat (Sarker 506 507 and Thorne, 2006; CEGIS, 2007), the use of variable time periods in this study means that a time-integrated measure of peak discharge (i.e. Q⁹⁵ exceedance) should also be 508 509 included in the analysis.

The CWT for each time period was scale-averaged into the following regular intervals: 0-10 km; 10-20 km; 20-30 km; 30-40 km; 40-50 km. The definite integral of each of the scale-averaged spectra for each time period was then quantified using the trapezoidal rule and plotted against the peak discharge measures determined from the available discharge records at the gauging station at Bahadurabad (Fig. 7).

515 At 0-10 km and 10-20 km scales, strong positive, exponential relationships are 516 seen to exist between the integral of the wave power spectrum and QMax / Q⁹⁵ 517 exceedance (Figs. 8 and 9). This indicates that as peak discharge increases, so too does

518 the total amount of bank retreat on the LHB of the Jamuna. It is interesting to note 519 that, in general, the strength of the relationships is stronger for maximum discharge than for Q⁹⁵ exceedance, suggesting that QMax may be a more useful peak discharge 520 521 parameter when attempting to predict bank retreat on the Jamuna. Positive 522 relationships also exist at larger scales, albeit with far lower integrated power spectrum 523 values, and lower Pearson coefficients. The positive relationships identified are not unexpected as the mean rate of bank retreat on the Jamuna is known to be related to 524 525 the magnitude of the largest monsoon flood (Sarker and Thorne, 2006; CEGIS, 2007). 526 However, the evidence that the strength of the relationship shows a consistent decrease 527 as scale increases is new and important knowledge. Overall reductions in the magnitude of the wavelet power spectrum integrals are accompanied by a reduction in R² values as 528 529 scale increases. For QMax the reduction is from 0.86 at 0-10 km scales to less than 0.5 at scales greater than 30 km. For Q^{95} exceedance the reduction is from ~0.5 at scales 530 531 less than 20 km to ~0.3 at scales greater than 40 km. Thus, the pattern is one in which 532 peak discharge is strongly related to the rate of bank retreat on the Jamuna, but only at 533 spatial scales of less than 20 km. As the scale of bank retreat increases the importance 534 and statistical significance of peak discharge as a main driver of bank retreat is 535 substantially reduced.

536 *3.4. Geomorphologic interpretation*

537 A number of key characteristics of bank line retreat on the LHB of the Jamuna 538 river have been identified through the CWT analyses presented. Whilst some can be 539 interpreted with reference to well-understood geomorphologic processes and 540 conventional geomorphological thinking, others are more difficult to interpret and further 541 work is required to provide an adequate geomorphological explanation.

542 The CWT sequence in Figure 5 shows that at different times in the Jamuna river 543 data series varying amounts of significant, short (2-16 km), moderate (16-32 km) and 544 long (>32 km) wave-period bank retreat are evident. In some years (e.g. 1996-97;

545 1997-98 and 1998-99), spatially-discrete regions of short wave-period bank retreat are 546 dominant and there is little evidence of significant retreat at longer periods. This 547 indicates that, at these times, the main mode of planform adjustment is local; through 548 the erosion of individual embayments. At other times (e.g. 1987-89; 1989-92; 1992-94; 549 1994-95 and 1995-96) the co-existence of significant regions of wavelet power at longer 550 wave-periods indicates that a regional mode of planform adjustment at or above the 551 scale of individual chars operates is also in operation. Indeed, in some years (1994-95 552 and 1995-96) significant regions of bank retreat at long wave-periods strongly suggest a 553 macro-scale model of adjustment that exceeds the scale of individual island chars.

554 The scale-averaged data presented in Figure 6 are more easily interpreted and 555 provide important geomorphological insights into the different characteristics of the 556 spatio-temporal patterns of bank retreat operating at different scales. At scales of 0-10 557 km the evidence of locally-persistent retreating banks, separated by stable and/or 558 transient reaches, corresponds well with the findings from other studies of local 559 embayment patterns on the Jamuna (Ellis, 1993). However, the precise reasons for the alternating pattern of stable / transient / eroding reaches at this scale are not fully 560 561 understood. At the scales at which gross planimetric control of bank retreat by quasi-562 stable island / node reaches (10-30 km) is thought to dominate (Coleman, 1969; Thorne 563 et al., 1993) there is relatively little evidence that the locations of significant bank 564 retreat can be mapped directly to the locations of island reaches. Indeed, Thorne et al., 565 (1993) identify seven separate island and nodal reaches located at roughly regular 566 downstream spacing throughout the 204 km study reach. However, no evidence of such 567 spacing in the pattern of significant bank retreat is evident in Figure 6, and only one 568 persistently stable reach is evident. This suggests that the importance of island and 569 nodal reaches on influencing the pattern and magnitude of bank retreat at large spatial 570 scales may be less important than previously thought. Instead, the downstream 571 propagating patterns of retreat observed correspond more closely to the influence of 572 sediment waves; the importance of which has been recognised in earlier studies (Takagi

573 et al., 2007). Indeed, they estimated a wavelength of 35 km and found the best 574 evidence for the propagation between 10 and 80 km downstream - approximately the 575 same wavelength and location of the patterns observed in Figure 6. Thus, additional 576 support is provided to the implication that propagating sediment waves are important 577 drivers of bank retreat patterns operating at scales of tens of kilometres on the Jamuna.

578 Relating the integral of a range of scale-averaged wavelet power spectra to maximum discharge and Q⁹⁵ exceedance, Figures 8 and 9 provide an important and 579 580 explicit confirmation of conventional geomorphological thinking about the importance 581 that can be ascribed to peak discharge as a driver of bank migration at different scales 582 (e.g. Hooke, 1980; Nanson and Hickin, 1986). On the LHB of the Jamuna the magnitude 583 of the peak discharge is strongly related to the integral of the wavelet power spectrum at 584 0-10 and 10-20 km scales and, hence, the magnitude of erosion at frequencies that 585 coincide with meander bend and embayment processes. At lower frequencies, where the 586 gross planimetric setting and regional factors such as the spatial variability of floodplain 587 substrate cohesiveness become important constraints on erosion, the relationship 588 between measures of peak discharge and wavelet power decreases.

589

4. Summary and conclusions

590 The Jamuna river provides an important venue for fundamental research on 591 process-form interactions in large-scale, complex fluvial systems. It is the epitome of a 592 wilful stream representing a complex, non-linear, dynamical system within which fluvial 593 processes, morphological responses and process-response feedback loops operate at 594 multiple scales of time and space. Past efforts at understanding this system have 595 generally focussed on studying the processes governing the evolution of individual 596 geomorphological units operating at a single scale within the channel, (e.g. anabranches, 597 bars, bends and bifurcations within the braided system). Yet, their value in 598 understanding how the system operates across its whole range of scales and periods of 599 adjustment are unavoidably limited by the approaches taken. As a consequence, the

multi-scale explanatory linkage between fluvial processes and channel evolution
envisaged first by Schumm and Lichty (1965) and later by Lane and Richards (1997)
remains poorly developed.

603 CWTs offer an important means by which key signals of planimetric change, in 604 the case of this study captured as sequences of bank migration spatial series, can be 605 localised not only in time, but also in space. This offers a powerful means of 606 characterising where, when and over what spatial scales change occurs. It thus 607 represents the first, vital step in determining a multi-scale, explanatory framework for 608 relating channel process and channel pattern evolution. In many cases, it will be possible to map the patterns observed to the results of past channel evolution process 609 610 studies. In this context, CWTs offer an important means of identifying spatio-temporal 611 patterns of bank retreat at different scales that can then be linked to fundamental 612 processes of channel adjustment and the findings of past research efforts. In other 613 cases, the patterns observed in the CWT will be more difficult to explain by conventional 614 geomorphological thinking. In these cases, CWTs offer an important means by which 615 new research directions can be identified and directed.

616 However, the outputs from a CWT are only as good as the input data. Indeed, 617 the selection of a planimetric signal capable of providing an adequate characterisation of 618 the evolutionary processes of interest is critical. For example, whilst sequences of bank 619 line retreat series offer important insights into erosion processes, they provide no 620 information about the temporal, spatial and frequency scales at which deposition occurs. 621 This means that an important component of channel change processes are 622 uncharacterised and the links between depositional and erosional channel forms can only 623 be surmised; not explicitly demonstrated. Similarly, by focussing on a single bank line, 624 only half of the channel's erosion response is characterised. Consequently, determining 625 the optimum set of channel response signals to which CWT should be applied in order to 626 gain a more holistic characterisation of the patterns of channel evolution is a pressing 627 research need. Also of importance is the quality and comprehensiveness of the data

628 used. Where the length of the data series is short, the relatively large size of the COI 629 will limit the ability to localise low frequency responses in the data. The downstream 630 resolution of the data will ultimately determine the degree to which high frequency 631 responses can be localised. Similarly, the length of time over which the signal is 632 measured, relative to the return period of the fluvial processes responsible for those 633 changes, may reduce the magnitude of certain responses when the data are converted 634 to annual rates. This in turn will reduce their wavelet power in the CWT and may result 635 in their significance being underestimated.

To conclude, the results presented in the paper represent an early, exploratory investigation of the usefulness of wavelet transformation of signals of planimetric change in complex river systems. Considerable potential for the technique is evident, however numerous questions remain and both wavelet analysis in general and CWT in particular offer considerable opportunities for fruitful future research efforts in unravelling river pattern change.

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Image Date	Sensor	Resolution	Acquisition Date	Q at Bahadurabad (m³ sec⁻¹)
1987	Landsat MSS	80m x 80m	7 Feb	4000
1989	Landsat TM	30m x 30m	28 Feb	6070
1992	Landsat TM	30m x 30m	8 Mar	4660
1994	Landsat TM	30m x 30m	25 Jan	5070
1995	Landsat TM	30m x 30m	28 Jan	4550
1996	Landsat TM	30m x 30m	31 Jan	4680
1997	Landsat TM	30m x 30m	18 Feb	No Data
1998	Landsat TM	30m x 30m	5 Feb	3710
1999	Landsat TM	30m x 30m	23 Jan	4830

Table 1. Images used to derive the bank line migration series.

Downstream	Reach	Description
Distance	Character	
0 - 8 km	COI	Falls within the COI.
8 - 57 km	Transient	Reach contains significant wavelet power of low,
	and	moderate and high magnitudes, but which is
	Variable	temporally and spatially transient. Bank retreatis
	Magnitude	evident at scales of 0-10 km, but its location and
	-	magnitude exhibits a high degree of temporal
		variability. There is strong evidence of downstream
		migration of bank line retreat.
57 - 65 km	Low	Little or no significant wavelet power at any time
	Magnitude	period. There is little evidence of bank retreat
	- J	operating at scales between 0-10 km.
65 - 85 km	Persistent	Reach has a consistent spatial and temporal pattern of
	and High	significant wavelet power of medium and high
	Magnitude	magnitude. Substantive bank retreat at scales of 0-10
	gritter	km is occurring throughout the study period.
85 - 120 km	Transient	Reach contains significant wavelet power of low and
00 120 1411	and	moderate magnitudes and which is temporally and
	Moderate	spatially transient. Moderate bank retreat is occurring
	Magnitude	at scales of 0-10 km. There is evidence of
	riagineade	downstream migration of bank line retreat.
120 - 135 km	Low	Little or no significant wavelet power at any time
120 135 Kill	Magnitude	period. There is little evidence of bank retreat
	Magnitude	operating at scales between 0-10 km.
135 - 155 km	Persistent	Reach has a consistent spatial and temporal pattern of
155 155 Kill	and High	significant wavelet power of moderate and high
	Magnitude	magnitudes. Substantive bank retreat at scales of 0-
	Magnitude	10 km is occurring throughout the study period.
155 - 164 km	Low	Little or no significant wavelet power at any time
105 104 KIII	Magnitude	period. There is little evidence of bank retreat
	magnitude	operating at scales between 0-10 km. This stable
		reach is probably a result of the guide bunds of the
		Jamuna Bridge.
164 - 196 km	Transient	Reach contains significant wavelet power of low,
104 190 KIII	and	moderate and high magnitudes, but it is temporally
	Variable	and spatially transient. Bank retreat is occurring at
	Magnitude	· · · ·
	maynitude	scales of 0-10 km, but its location and magnitude
		exhibit a high degree of temporal variability. There is
		strong evidence of downstream migration of bank line
<u></u>	<u> </u>	retreat
COI	COI	Falls within the COI.
196 - 204 km		

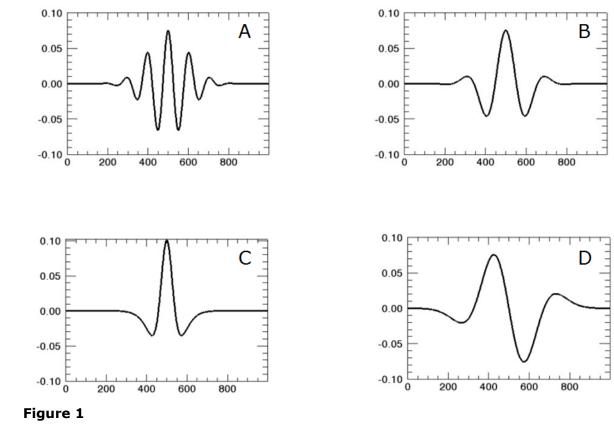
Table 2. Characteristic reach types at the 0-10 km scale band.

874	Table 3. Maximum daily mean discharge (QMax) and the number of days exceeding the
875	total study period 95 th percentile discharge (Q^{95}) for each time period.

Time Period	QMax recorded at Bahadurabad (m ³ sec ⁻¹)	Days exceeding Q_{95}
1987-89	98,300	42
1989-92	84,100	60
1992-94	67,000	18
1994-95	40,900	0
1995-96	87,000	15
1996-97	83,800	21
1997-98	79,219	5
1998-99	103,128	48

- Figure 1. Mother wavelets for A. Morlet (*k*=6), B. Morlet (*k*=3), C. Paul (*k*=3) and D.
 DOG (*k*=3).
- **Figure 2.** The location of the study reach showing the confluence with the Teesta River (A), the location of the Bahadurabad gauging station (B), the approximate position of the guide bunds of the Jamuna Bridge (C) and the confluence with the Ganges River (D). Note that the main channel is orientated due north-south.
- Figure 3. The temporal sequence of 1987-1999 bank line retreat series for the Jamunariver LHB.
- **Figure 4.** CWT for 1998-99 bank line retreat series using Morlet k = 3 and k = 6, Paul k 888 = 3 and DoG k = 3 mother wavelets. The bold line shows the 95% confidence level (i.e. 889 regions within the line have significant wave power over a background red noise 890 spectrum $\alpha = 0.72$). The thin dashed line shows the COI. Regions within the COI should 891 be interpreted with caution.
- **Figure 5.** CWT power spectra (Morlet k=3 wavelet function) for bank line retreat series for all time periods. The bold line shows the 95% confidence level (i.e. regions within the line have significant wave power over a background red noise spectrum). The thin line indicates the COI. Regions within the COI should be interpreted with caution.
- Figure 6. Scale averaged wavelet power spectra for 0-10 km and 10-30 km scale bands.
 Data are plotted as multiples of the wavelet power value equal to the 95% confidence
 level. The analysis ignores the regions of the plots where variation of wavelet power
 occurs within the cone of influence (labelled COI).
- Figure 7. Daily mean discharge at Bahadurabad for the 1987-1999 study period. Theportion of the hydrograph within each time period is indicated.
- 902Figure 8. The relationship between Q_{Max} and the integral of the scale-averaged wavelet903power spectrum for scale bands 0-10 km, 10-20 km, 20-30 km, 30-40 km and 40-50 km.
- Figure 9. The relationship between the number of days for which flow exceeds the 95th
 percentile for the study period (Q₉₅) and the integral of the scale-averaged wavelet
 power spectrum for scale bands 0-10 km, 10-20 km, 20-30 km, 30-40 km and 40-50 km.
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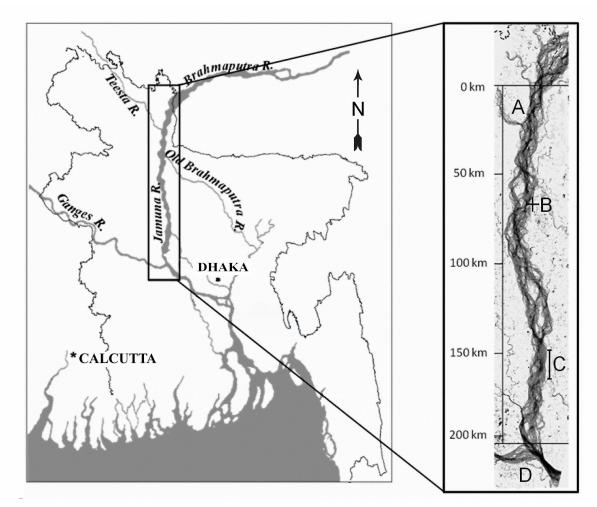
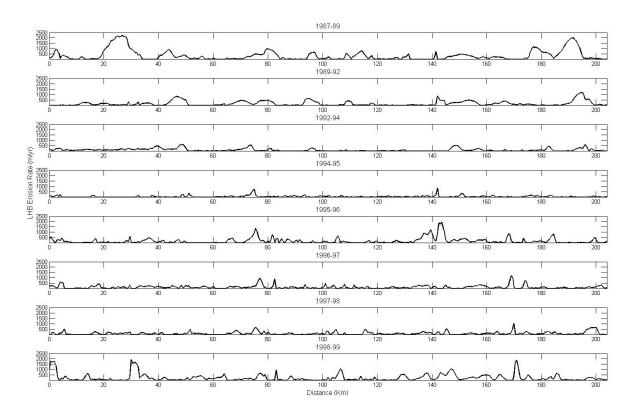




Figure 2



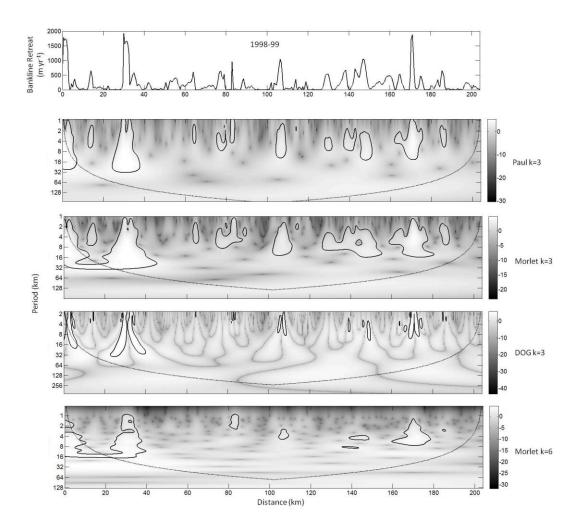


Figure 4

