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Tracking Mass-Wasting Sediments in the Large Brahmaputra River System

Abstract

The impact of mass-wasting events on river systems is typically studied in the immediate aftermath and within close proximity to the source area. However, basin-wide hydro-geomorphic responses in large river systems, beyond the immediate post-disaster context, remain poorly understood. In this study, we examine the Eastern Himalayan Syntaxis in the Brahmaputra basin, where the Sendongpu glacier valley has undergone rapid erosion due to major mass-wasting events in 2017, 2018 and 2021. A nearby large-scale event also occurred in 2000 along the Yigong River. We assess the river's response to these events using satellite-derived water turbidity indices, water surface elevation, sandbar area and flood extent, at the finest temporal and spatial resolution available through Google Earth Engine's database. Our time-series analysis reveals that, following the 2017 event, fine suspended sediment signals can be traced more than 1000 km downstream to the delta, while coarser sediments primarily affect river morphology within approximately 100 km of the mountain front. In this zone, we observed a notable increase in sandbar area and water surface elevation, indicating extensive deposition and channel infilling. Comparison with the 2000 Yigong outburst highlights the differing geomorphic effects of rapid, high-magnitude events versus slower, more sustained sediment inputs. These findings highlight the need to integrate mass-wasting-driven sediment processes into flood risk assessments and hydropower planning in Himalayan river systems, particularly as such landscapes are increasingly subject to both natural and anthropogenic pressures.

Keywords: Sendongpu, Yigong, Brahmaputra, Large rivers, Mass wasting, Google Earth Engine, GLOFs

1 Introduction

Mass-wasting events generate large volumes of sediment that can profoundly impact both society and the environment (Carrivick and Tweed, 2016; Taylor et al., 2023; Froude and Petley, 2018; Fidan et al., 2024; Sharma et al., 2023). Their effects are typically twofold: (i) direct impacts, where the large sediment surge damages infrastructure, dams, and settlements (e.g., Shugar et al., 2021; Sattar et al., 2025; Li et al., 2022; Zhang et al., 2023), and (ii) indirect impacts, where the influx of sediment reshapes river corridors, modifies channel morphology, and elevates flood risk (Korup, 2004; Li et al., 2020; Sofia and Nikolopoulos, 2020; Ahrendt et al., 2022; Vázquez-Tarrío et al., 2024).

One of the largest recorded mass-wasting events as a co-seismic landslide occurred after the 1950 Assam earthquake, which released $47000 \times 10^6 \text{ m}^3$ of sediment, causing prolonged channel aggradation and widespread flooding along the large Brahmaputra River, lasting for decades (Goswami, 1985; Keefer, 1999). The 2021 Chamoli disaster in the upper Ganga basin, with a volume of $23 \times 10^6 \text{ m}^3$, severely damaged hydropower infrastructure (Shugar et al., 2021). More recently, the 2023 Sikkim flood, triggered by the collapse of $14.7 \times 10^6 \text{ m}^3$ of frozen moraine into a glacier lake, caused an outburst flood that mobilized $270 \times 10^6 \text{ m}^3$ of sediment, devastating infrastructure along the Teesta River (Sattar et al., 2025).

Mass-wasting can be triggered by a range of drivers, including meteorological extremes, earthquakes, wildfires, land use changes, and vegetation disturbances (Yanites et al., 2025; Jones et al., 2021; Kashyap et al., 2025). These events also take multiple forms. In glaciated landscapes, ice-rock avalanches, glacier and landslide outburst floods are common, whereas in tectonically active mountain regions, co-seismic landslides and subsequent outburst events often dominate. Extreme precipitation frequently generates widespread landslides (e.g. Rinaldi et al., 2016), delivering large volumes of sediment to river systems. For the purposes of this study, we broadly define mass-wasting as any event capable of supplying anomalously high sediment loads to the fluvial system, whether or not these sediments are ultimately transported downstream.

Once sediments are released, they enter the fluvial system and are transported downstream over spatial and temporal scales that vary widely depending on hydrometeorological conditions (Graf et al., 2024; Bailey et al., 2025; Gayer et al., 2025; Zhang et al., 2016; Wang et al., 2015), topographical or morphological conditions (Croissant et al., 2017; Dai et al., 2021; Zhang et al., 2016; Wang et al., 2015), grain size of the deposits (Fordham et al., 2023; Wang et al., 2015; ?) and connectivity of the hillslope to a fluvial network (Graf et al., 2024; Yanites et al., 2010). Finer

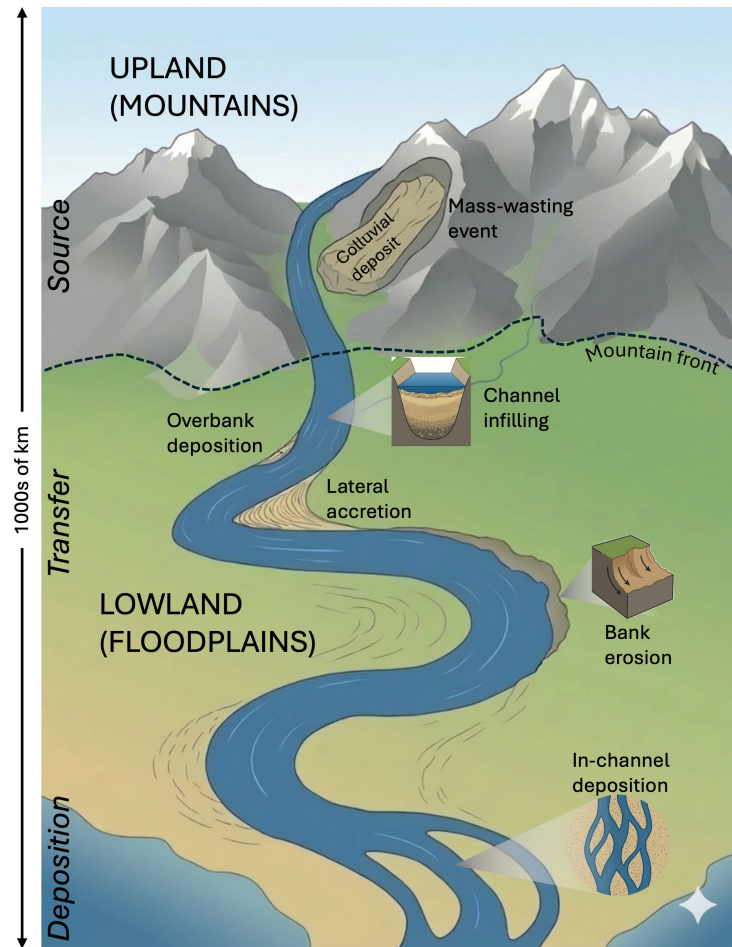


Figure 1: Conceptual diagram showing the transport of mass-wasting sediments and geomorphic alternation in the large river course. Gemini (version 3.00) was used to improve the artistic design of this figure.

53 sediments in suspension would travel further and faster than the coarse sediments.

54 The evacuation of mass-wasting sediments typically occurs in two stages: first, the transfer of
 55 the material from source (hillslopes or glacier valleys) to river channels; and second, the downstream
 56 transport of these sediments once they enter the fluvial network (Croissant et al., 2017). Along this
 57 pathway, various geomorphic units act as temporary storage zones (Figure 1), influencing evacuation
 58 times. For example, freshly failed sediment may be stored temporarily as colluvial deposits on hill-
 59 slopes, only to be mobilized later under favorable hydrometeorological conditions (Figure 1). Along
 60 the routing pathway, sediment can be stored in overbank deposits, reworked into in-channel bars,
 61 or contribute to channel infilling. Tracing *how far* these sediments travel and *how long* they persist
 62 is therefore essential for understanding the compounded risks associated with mass-wasting (?).

63 To address the question of time scales (how long) mass-wasting sediments remain in transit,
64 several studies report highly variable timescales ranging from days (Yunus et al., 2025) to years
65 (e.g. Westoby et al., 2023; Croissant et al., 2017; Gayer et al., 2025; Hovius et al., 2011; ?), and
66 in some cases, decades (e.g. ?Major et al., 2018; DeLisle et al., 2022; Li et al., 2025; Xiong et al.,
67 2022). However, these studies are mostly limited to a smaller spatial scale (tens to a few hundreds
68 of kilometers). Therefore, an equally important question concerns the spatial extent (how far) of
69 transport. There is a little understanding of geomorphological response (if any) in large river basins
70 (area > 10⁵ km², length > 10³ km). Large river basins integrate highly diverse sediment sources,
71 rainfall regimes, and source–sink dynamics (Ashworth and Lewin, 2012; Tandon and Sinha, 2022;
72 Jain et al., 2025). This raises a fundamental question: can localized mass-wasting events, that impact
73 only a very small portion of a basin’s area, leave detectable signatures of sediment dispersal and
74 geomorphic change across an entire large river course?

75 With the advent of cloud-based platforms such as Google Earth Engine, large-scale tracking of
76 mass-wasting signatures is now feasible. Because these events are both short-lived and spatially
77 localized, high spatial and temporal resolution imagery is critical to track the change and response
78 downstream. Available datasets in Google Earth Engine make it possible. While field-based surveys,
79 digital elevation differencing, and numerical models (e.g. in Bennett et al., 2025; Yang et al., 2022)
80 provide volumetric estimates and mechanistic insights of sediment transport, they remain logisti-
81 cally and computationally constrained at basin scales, particularly in large rivers. Instead, remote
82 sensing proxies, such as changes in sandbar area, turbidity, and floodwater extent, offer a consis-
83 tent, scalable means to assess fluvial response to mass-wasting events across both space and time.

84 In the Eastern Himalaya, erosion rates are among the highest globally (Kääb and Girod, 2023).
85 Over the last 7–8 years, the Sendongpu glacier alone has eroded approximately 700 × 10⁶ m³ (1400
86 Mt) of sediment (Li et al., 2024), nearly three times the Brahmaputra River’s annual sediment load
87 of approximately 520 Mt (Milliman and Farnsworth, 2011). Moreover, this region also has a history
88 of megafloods, where outburst floods released discharges and sediment loads several orders of
89 magnitude greater than the present river flow (Lang et al., 2013; Panda et al., 2020; Turzewski et al.,
90 2020). This raises critical questions about the fate of such massive sediment inputs. This study
91 evaluates the *spatial extent* and *persistence* of fluvial response to large-magnitude mass-wasting
92 events in a large river system. We study the mass-wasting events in Sendongpu and the nearby
93 Yigong River, examining the resulting morphological and sedimentary signals in the downstream
94 Large Brahmaputra River across large spatial (1000s of km) and temporal (years) scales, extending

95 our understanding of processes beyond the immediate, localized post-disaster context.

96 **2 Study site and mass-wasting events**

97 The Brahmaputra is a large river basin with an area of about 580,000 km², and a trunk length of
98 nearly 3,000 km (Figure 2). It originates on the Tibetan Plateau, where it is known as the Yarlung
99 River. After entering the Himalaya, it flows as the Siang through steep topography until reaching the
100 mountain front near Pasighat. Downstream of Pasighat, the river adopts a braided form and flows
101 as the Brahmaputra before draining into the Bay of Bengal. The braided belt width reaches upto 20
102 km with a braiding index mostly varying between 5 to 15 (Sarma and Acharjee, 2018). The braiding
103 pattern has the presence of sandbars, low-flow channels, chutes, cutoffs and vegetated landforms
104 (Nandi et al., 2022). The Brahmaputra is a sediment-laden river with an annual sediment load of 500
105 to 1000 ×10⁶ tonnes/year (Islam et al., 1999; Galy and France-Lanord, 2001).

106 A key geomorphic feature of the basin is the Eastern Syntaxis, where the Yarlung River makes
107 a dramatic 180° turn. This region experiences some of the highest erosion rates in the Himalaya
108 (4–21 mm/year) (Lupker et al., 2017; Garzanti et al., 2004; Stewart et al., 2008). In recent decades,
109 it has also been the site of four major mass-wasting events: the 2000 Yigong event and the 2017,
110 2018, and 2021 Sendongpu events. The Sendongpu valley lies within the Eastern Syntaxis, while the
111 Yigong site, about 50 km away, is located along the Yigong River, a tributary that joins the Yarlung
112 downstream of Sendongpu.

113 A summary of the mass-wasting events in the Sendongpu and Yigong is provided in SI Table S1.
114 The events are described in detail in the following subsections.

115 **2.1 Yigong event (2000)**

116 The 2000 Yigong event is a classic example of a compounding outburst flood. Located approximately
117 50 km from the site of the Sendongpu events, this disaster began on April 9, 2000, with a massive
118 landslide of approximately 100 × 10⁶ m³ (Guo et al., 2023; Dong et al., 2024). The landslide evolved
119 into a rock avalanche that blocked the Yigong River, forming a debris dam with an estimated volume
120 of 300 × 10⁶ m³ (Guo et al., 2023). The blockage created a lake with a volume of approximately 3000
121 × 10⁶ m³ at the time of breaching (Guo et al., 2023). On June 12, 2000, the dam burst suddenly,
122 draining the lake completely and causing extensive damage to downstream areas. In addition, Dong
123 et al. (2024) estimated that an additional 260 × 10⁶ m³ of sediments were eroded from the valley

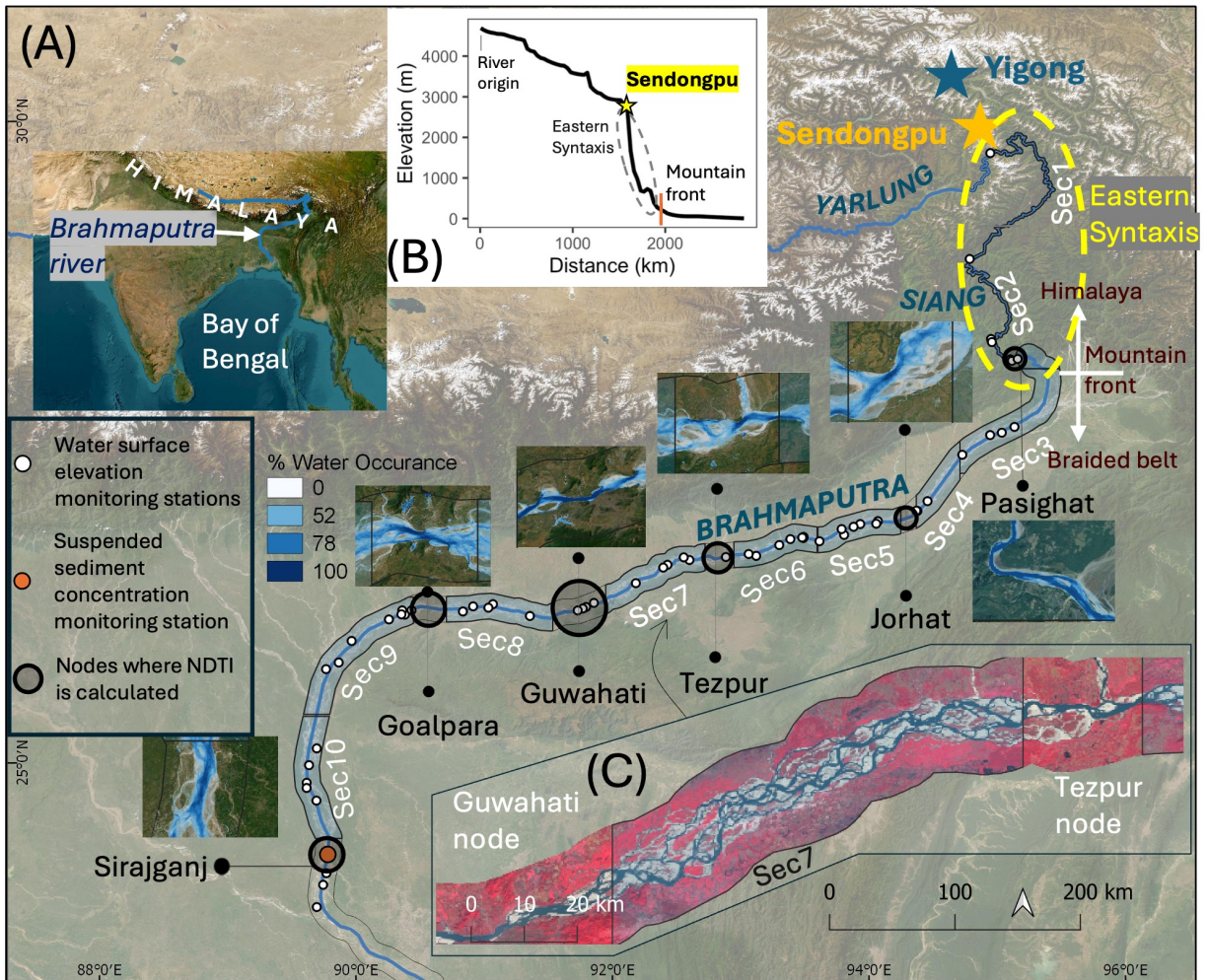


Figure 2: (A) Location of the Yarlung-Siang-Brahmaputra River in the eastern Himalaya, highlighting the studied mass-wasting sites, Yigong and Sendongpu. The river is divided into ten sections (Sec1 to Sec10), each defined by a 20 km buffer around the river centerline. River node locations, derived from the JRC-Global Surface Water dataset (Pekel et al., 2016), are shown on the water occurrence raster and indicate locations of minimal river width and maximum water occurrence probability. Water surface elevation and suspended sediment concentration time series were obtained from the monitoring datasets of Schwatke et al. (DAHITI, 2015) and Dethier et al. (2022), respectively. (B) River long profile along the Yarlung-Siang-Brahmaputra. (C) Below the mountain front, the river transitions to a braided form, shown in false color composite for Sec7. Field photographs of certain locations are shown in SI Figure S1. Basemap source: Esri, Maxar, Earthstar Geographics, and the GIS User Community.

124 floor and concurrent landslides were triggered due to the outburst as far as 80 km downstream of
125 the outburst flood location.

126 **2.2 Sendongpu events (2017-2021)**

127 Sedongpu has a history of mass wasting events (Kääb and Girod, 2023). In this study, we consider
128 three events, one in each of 2017, 2018 and 2021. During 2017, there were ice-rock avalanche and
129 debris flows from 20 Oct 2017 to Jan 2018. The total volume of material that detached from the
130 hillslope during this time was $50 \times 10^6 \text{ m}^3$. On 18 Nov 2017, a Mw 6.4 earthquake also struck the
131 region (Figure 3). This earthquake triggered 1816 landslides in the upper section of the Siang River
132 (Sec 1 in Figure 2) (Hu et al., 2018, 2019). The average volume of these landslides was estimated at
133 $14 \times 10^6 \text{ m}^3$ (Zhao et al., 2021). Occasionally, the debris deposits blocked the river in this period,
134 which eventually breached or overtopped (Hu et al., 2018; Zhao et al., 2019, 2021).

135 Following the series of events in 2017, significant changes were observed on the glaciers of the
136 Sendongpu valley (Kääb et al., 2021). Eventually, a large glacier detachment occurred a year later
137 in 2018 (October 17) at the Sendongpu glacier tongue. The glacier detachment mobilized a total
138 of $180 \times 10^6 \text{ m}^3$ of sediment from the glacier tongue. The river blockage caused by these events
139 resulted in the formation of a lake with a volume of approximately $490 \times 10^6 \text{ m}^3$. The debris fan that
140 formed the dam had a crest length of 300 m along the river, a base length of 1.9 km and a height
141 of 79 m (Chen et al., 2020). The dam stored an estimated $35 \times 10^6 \text{ m}^3$ of sediment, accounting for
142 approximately 20% of the total sediment mobilized from the detachment of the glacier tongue (Li
143 et al., 2024).

144 The 2021 Sendongpu events were among the most intense and significant in the region. On
145 March 22, 2021, a massive ice-rock avalanche occurred, mobilizing approximately $50 \times 10^6 \text{ m}^3$ of
146 material from the eastern flank of the Sendongpu glacier (Kääb and Girod, 2023). Subsequently,
147 between April and September 2021, an additional $279 \times 10^6 \text{ m}^3$ of sediment was mobilized from
148 the glacier tongue area, where similar events had occurred in 2018.

149 Unlike earlier events, those in 2021 did not cause a significant river blockage, allowing the river
150 to continue flowing uninterrupted. However, the exact quantity of sediment that entered the river
151 and was transported downstream remains unquantified. Nevertheless, studies by Zhao et al. (2022)
152 and Kääb and Girod (2023) suggest far-reaching downstream transport of the mobilized sediments.

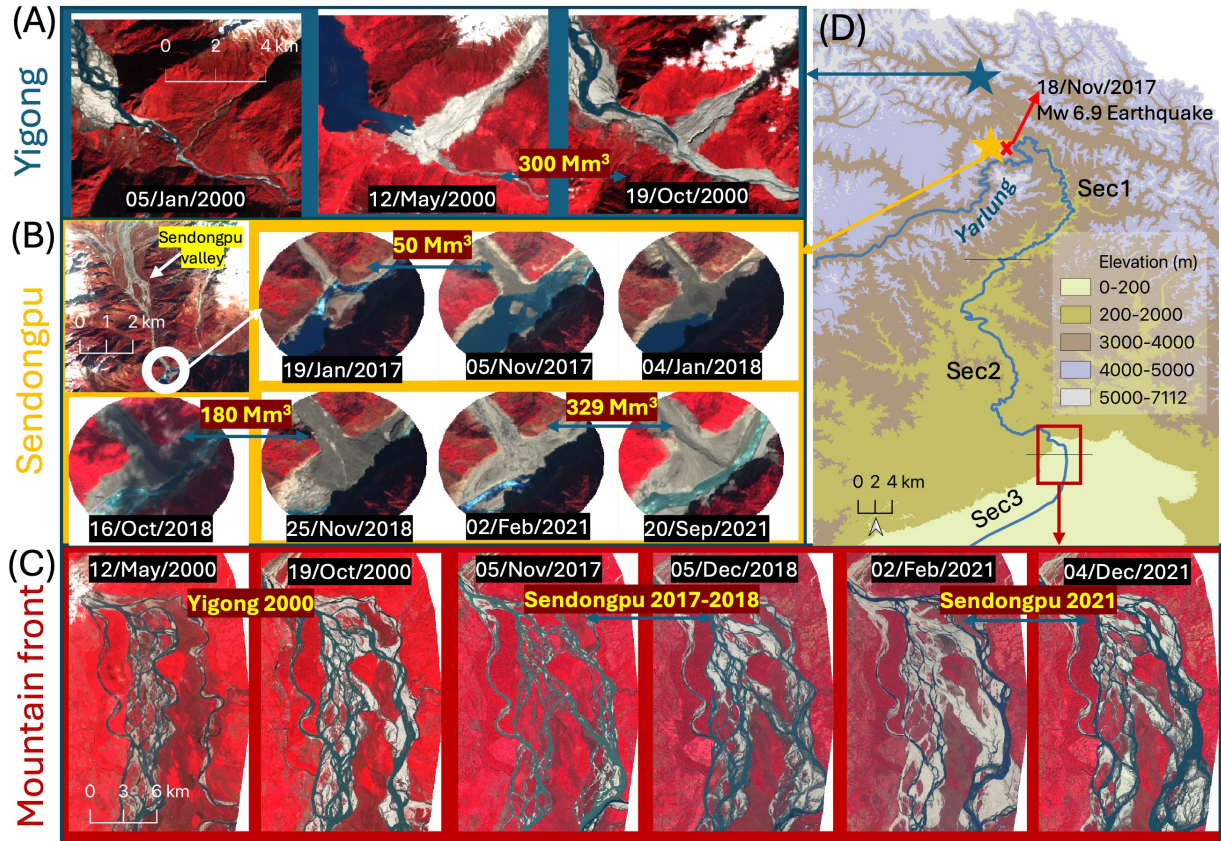


Figure 3: Location of the Yigong (A) and Sendongpu (B) mass-wasting site with the timeline and magnitude of mass-wasting events in 2000, 2017, 2018 and 2021. Sediment volumes generated by these events are shown in yellow. Panel C shows the river planform response downstream of the mountain front (part of Sec3) following the Yigong and Sendongpu events. Panel D shows the elevation map of the study region. All optical satellite images are in false color composite.

Table 1: Summary of the derived dataset and its attributes used in this study.

Parameter	Source	Spatial Resolution	Spatial Coverage	Temporal Resolution	Temporal Coverage
Normalize difference turbidity index	Landsat5, Landsat8	30 m	6 time series zones along the Brahmaputra	15 days	1988-2024
% Sandbar area	Dynamic world	10 m	10 sections along the Yarlung-Siang-Brahmaputra	5-10 days	2017-2024
Water surface elevation	DAHITI	-	59 time series locations along the 1000 km Yarlung-Siang-Brahmaputra	10-35 days	2002-2024
Suspended sediment concentration	Dethier et al., (2022)	-	1 time series location in delta	15 days	1988-2020
Optical imagery	Landsat 7, Sentinel 2	30 m, 10 m	Stretch of Yarlung-Siang-Brahmaputra	5, 15 days	2000, 2017-2024
Microwave imagery	Sentinel 1	10 m	Stretch of Yarlung-Siang-Brahmaputra	12 days	2017-2024
Rainfall	CHIRPS	5.4 km	Basin of Yarlung-Siang	5 days	2000-2024

3 Data and methods

In this study, multiple proxies were employed to investigate changes in the river planform, water turbidity/suspended sediment concentration and river water surface elevation in relation to mass-wasting events. These parameters were analyzed across the entire Yarlung-Siang-Brahmaputra stretch of approximately 1230 km at the finest spatial (10-30 m) and temporal resolution available (5-30 days). A summary of the datasets used, along with their spatial and temporal resolutions and coverage, is provided in Table 1.

To assess changes following mass-wasting events, we focused on suspended sediment, using the satellite-derived turbidity index as a primary proxy, complemented by suspended sediment concentration data at a single location in Sirajganj from [Dethier et al. \(2022\)](#) (Figure 2). Additionally, we examined changes in the river planform by utilizing the Dynamic World global land use and land cover dataset ([Brown et al., 2022](#)). As channel forms are expected to change due to these events, we also analyzed variations in water surface elevation. These changes serve as a proxy for channel infilling, which typically leads to an increase in water surface elevation.

3.1 River water turbidity

The Normalized Difference Turbidity Index (NDTI) is a remote sensing-derived index used to measure water turbidity by leveraging the difference in reflectance between specific spectral bands sensitive to sediment concentration in water. The NDTI is calculated using the red and green spectral

171 bands as:

$$NDTI = \frac{Red - Green}{Red + Green}$$

172 The resulting NDTI values range from -1 to 1, where higher values indicate greater turbidity
173 (Lacaux et al., 2007). Similarly, the Modified Normalized Difference Water Index (MNDWI) is used
174 to differentiate water pixels from non-water pixels. It is calculated as:

$$MNDWI = \frac{Green - SWIR}{Green + SWIR}$$

175 where SWIR refers to the shortwave infrared band. MNDWI values range from -1 to 1, with
176 negative values indicating water pixels (Xu, 2006).

177 To capture turbidity dynamics, the NDTI was calculated at six nodes, namely Pasighat, Jorhat,
178 Tezpur, Guwahati, Goalpara and Sirajganj, along the main trunk of the river (Figure 2). These nodes
179 were selected based on their geomorphic characteristics, where the river exhibits minimum width
180 and predominantly follows a single-channel flow pattern with high water occurrence probabilities
181 (Figure 2 and SI Figures S1b, S1c, S1e). This approach ensured we selected the areas where the
182 water flux was concentrated within a narrow channel, increasing the likelihood of obtaining repre-
183 sentative turbidity signatures in the water crossing each cross-section. In contrast, sections with
184 braided channels were excluded due to the difficulty in identifying a representative channel.

185 The analysis was conducted using Google Earth Engine, a cloud-based geospatial processing
186 platform. Landsat 5 imagery was used to calculate the NDTI for the period 1984–2013 and Landsat
187 8 imagery was used for 2013–2024. The MNDWI was first applied to all images to delineate water
188 pixels, ensuring that the NDTI calculations were restricted exclusively to water-covered areas. For
189 each of the selected nodes, median NDTI values were computed and aggregated at a daily resolution
190 for cloud-free pixels. Since the median statistic was applied to the NDTI values of all water pixels,
191 no filtering based on pixel count was performed. This approach assumes spatial homogeneity in
192 water turbidity for the small area covered by each node. SI Figure S2 illustrates the pixel count for
193 each image used in the computation of median NDTI values. The mean standard deviation of NDTI
194 values within a single image was approximately 0.03.

3.2 Planform analysis

Planform change analysis was conducted by estimating the exposed sandbar area in sections of the braided reach. We utilized the Dynamic World dataset from 2017 to 2024 available in Google Earth Engine. It is a global, high-resolution land cover dataset utilizing deep learning models applied to Sentinel-2 satellite imagery to provide near real-time dynamic classifications of land cover (Brown et al., 2022). It includes nine distinct land cover classes: water, trees, grass, flooded vegetation crops, shrubs, built-up areas, bare ground and snow/ice. The bare class corresponds to the exposed sandbar. We estimated the percent sandbar area in Google Earth Engine for ten sections each of approximately equal length, as shown in Figure 2. Only scenes with the highest pixel count (representing maximum spatial coverage) were selected for further analysis (black points in SI Figure S2B). Scenes with lower pixel counts due to partial coverage and cloud cover were excluded (shown in light gray in SI Figure S2B). Sandbar area proportion was given as:

$$\% \text{Sandbar area} = \text{Bare area pixel count} / \text{Total pixel count}.$$

The Dynamic World dataset has an overall classification accuracy of 73%, with the "bare ground" class achieving an accuracy of approximately 60% (Brown et al., 2022).

3.3 Water extent mapping using Synthetic Aperture Radar

To identify the potential effect of mass-wasting events on the flood area, we performed a temporal analysis of the maximum water extent in Sec3 (Figure 2) using Synthetic Aperture Radar (SAR) imagery acquired from the Sentinel-1 satellite on the Google Earth Engine platform. Since the maximum water extent occurs during the monsoon months when significant cloud cover limits the applicability of optical data, application of SAR based analysis gives an opportunity to get insights of flood area during the monsoon months.

The SAR images from Google Earth Engine provide normalized, calibrated and ortho-rectified backscatter coefficients (σ_o), which were converted to decibels (dB) using the equation $10 \log_{10} \sigma_o$. The backscatter coefficient indicates the proportion of radar energy reflected back to the antenna, influenced by factors such as dielectric constant, surface roughness, topography and local incidence angle. Water has a high dielectric constant, resulting in minimal backscatter compared to land surfaces. Smooth water surfaces produce significantly lower backscatter, whereas rougher surfaces exhibit slightly higher values, assuming other factors remain constant (Campbell and Wynne, 2011; Smith et al., 1996).

225 The images were obtained in Ground Range Detected mode, using the VH polarization band with
226 a spatial resolution of 30 m. In this analysis, we focused on the VH polarization band, as it is less
227 sensitive to river surface roughness compared to other bands (Dubey and Singh, 2021). To derive
228 the annual maximum water extent, we created a minimum-value composite of the image collection.
229 The minimum-value composite takes the minimum value of each pixel across all the images in the
230 image collection of each year, from January to December, which covers the monsoon months of
231 April to October. This approach assumes that pixels with the lowest backscatter values during the
232 year are most likely water pixels, as inundated areas exhibit minimal radar backscatter. This analysis
233 was undertaken for the 2015 to 2024 period, and include all available SAR images. Ultimately,
234 this yielded ten images of minimum-value composite corresponding to each year, and they, in turn,
235 indicated the maximum water extent most likely corresponding to the monsoon months of each
236 year.

237 It should be noted that Sentinel-1 has a nominal revisit time of 12 days, with more frequent
238 observations in overlapping swaths. However, since SAR acquisitions are not always synchronous
239 with the true peak of river discharge, the resulting composites may underestimate the absolute
240 flood extent in any given year. Nevertheless, the consistent use of minimum composites across
241 years allows for the detection of relative trends. Thus, while the method may not capture exact
242 peak flood extents, it can effectively reveal persistent, systematic changes in maximum water extent
243 associated with sediment-driven morphological alterations during the monsoon months.

244 **3.4 Hydrometric data and other satellite products**

245 We utilized data from the DAHITI (Database for Hydrological Time Series of Inland Waters) plat-
246 form to analyze changes in water surface elevation along the Brahmaputra River (Schwatke et al.,
247 2015). DAHITI employs satellite altimetry data from multiple missions (e.g., Jason-3 and Sentinel-3)
248 to generate water level time series for inland water bodies. For this study, water surface elevation
249 time series were obtained at 59 locations along the main trunk of the Brahmaputra, spanning from
250 Sendongpu to Sirajganj (Figure 2). The temporal frequency of the observation is 10-35 days. The
251 median error of each of these locations varies from 0.001 m to 0.19 m. To ensure comparability,
252 water surface elevation at each location was normalized to local minimum and maximum values.

253 For visual inspection of the planform change, we collected optical satellite images from the
254 Sentinel-2: MSI (Multi Spectral Instrument), specifically for the red, green and near-infrared bands,
255 and a time series of suspended sediment concentration was sourced from Dethier et al. (2022) at

256 the Sirajganj (Figure 2). Rainfall data at a 5-day time interval was provided by the Climate Hazards
257 Group InfraRed Precipitation with Station (CHIRPS) dataset (Funk et al., 2015), and was accessed via
258 GEE platform and an annual composite of mean monsoon rainfall was derived to identify potential
259 extreme meteorological controls on the geomorphological response.

260 3.5 Statistical analysis

261 To detect the significance of the change in parameters, we performed non-parametric statistical
262 tests due to the lack of normality in the data distributions. Each parameter was categorized into
263 test groups after visual inspection of the long time series. For each parameter, we first applied the
264 Kruskal-Wallis test, a rank-based method, to assess the overall significance of differences across
265 groups (Kruskal and Wallis, 1952). If the Kruskal-Wallis test indicated significant differences ($p <$
266 0.05), we conducted Dunn's post-hoc test with Bonferroni correction to identify specific pairwise
267 differences (Dunn, 1964). The details of these tests are presented in SI Section SS1 and SI Figures
268 S10, S11.

269 4 Results

270 4.1 Suspended sediment dynamics

271 Typically, NDTI values during the non-monsoon (October-April) season are low due to low sediment
272 concentration. However, mass-wasting events of 2017 in Sendongpu and co-seismic landslides led
273 to the increase in non-monsoon NDTI values at Pasighat to the levels comparable to those observed
274 during the monsoon months (April-September) (Figures 4A and SI Figure S3). Figure 3C and SI
275 Figure S4 illustrate the river channel near Pasighat with visibly turbid water on November 5, 2017,
276 compared to other times. Local media reports (SANDRP, 2017; Times of India, 2017) indicated
277 turbidity levels as high as 425 NTU on November 27, 2017 in Siang, significantly exceeding the
278 normal range of 12–15 NTU for that time of year.

279 The 2017-2018 non-monsoon NDTI values were significantly ($p < 0.05$) higher at Pasighat than
280 all other non-monsoon periods as a whole post-2012. Downstream nodes also recorded elevated
281 turbidity during the same period, though the intensity diminished further downstream at Sirajganj
282 ($p = 0.08$). However, post-monsoon suspended sediment concentrations at Sirajganj remained sig-
283 nificantly higher ($p < 0.05$) from 2017–2020 compared to the 2013–2016 period (Figure 4B).

284 Similarly, non-monsoon NDTI values in 2018–2019 remained significantly elevated at Pasighat,
285 Jorhat and Tezpur ($p < 0.05$) but were not statistically significant at Guwahati, Goalpara and Siraj-
286 ganj, suggesting a gradual attenuation of the sediment signature downstream in later years. During
287 the 2021 monsoon, when much of the sediment from the 2021 mass-wasting events was trans-
288 ported, turbidity levels were significantly ($p < 0.05$) elevated at Pasighat and Jorhat compared to
289 the monsoon period of 2013-2024 as a whole (Figure 4A). [Zhao et al. \(2022\)](#) also mentioned a high
290 amount of turbidity at 200 km downstream of Sendongpu. However, neither the 2018 nor the
291 2021 events caused turbidity increases as dramatic as those observed after the 2017 event. Ad-
292 ditionally, the 2000 Yigong event is distinctly identifiable as an outlier in the suspended sediment
293 concentration records at Sirajganj.

294 4.2 Water surface elevation

295 Figure 5 illustrates water surface elevation variations in and around Sec3. The normalized water sur-
296 face elevation for the entire Yarlung-Siang-Brahmaputra stretch is provided in SI Figure S5. Across
297 all sections and nodes, a consistent elevation pattern is observed, except for two data locations in
298 Sec3 (locations 7 and 10 in Figure 5), which exhibit notable variations. These two locations are the
299 only ones within Sec3 with time series data extending beyond 2016. Both show a significant in-
300 creasing trend ($p < 0.05$, SI Figure S10) in water surface elevation, especially during non-monsoon
301 months (Oct-Mar). The temporal alignment of these changes with upstream mass-wasting events
302 (marked by red stars in Figure 5) is particularly notable.

303 The water surface elevation increases during non-monsoon months in Sec3 suggest channel
304 infilling as the likely mechanism. While monsoonal water surface elevation fluctuate annually, they
305 remain constrained by an upper limit as excess water spills onto the floodplain/sandbars. In contrast,
306 non-monsoon low water surface elevation is more sensitive to external controls as they fluctuate
307 within the channel boundary.

308 Sec3 extends approximately 100 km downstream of the mountain front (near location 6 in Figure
309 5), beyond which, in Sec4 and Sec2, no significant ($p > 0.05$) elevation changes are detected. This
310 indicates that planform changes are largely confined to the first 100 km downstream of the mountain
311 front (location 6 in Figure 5).

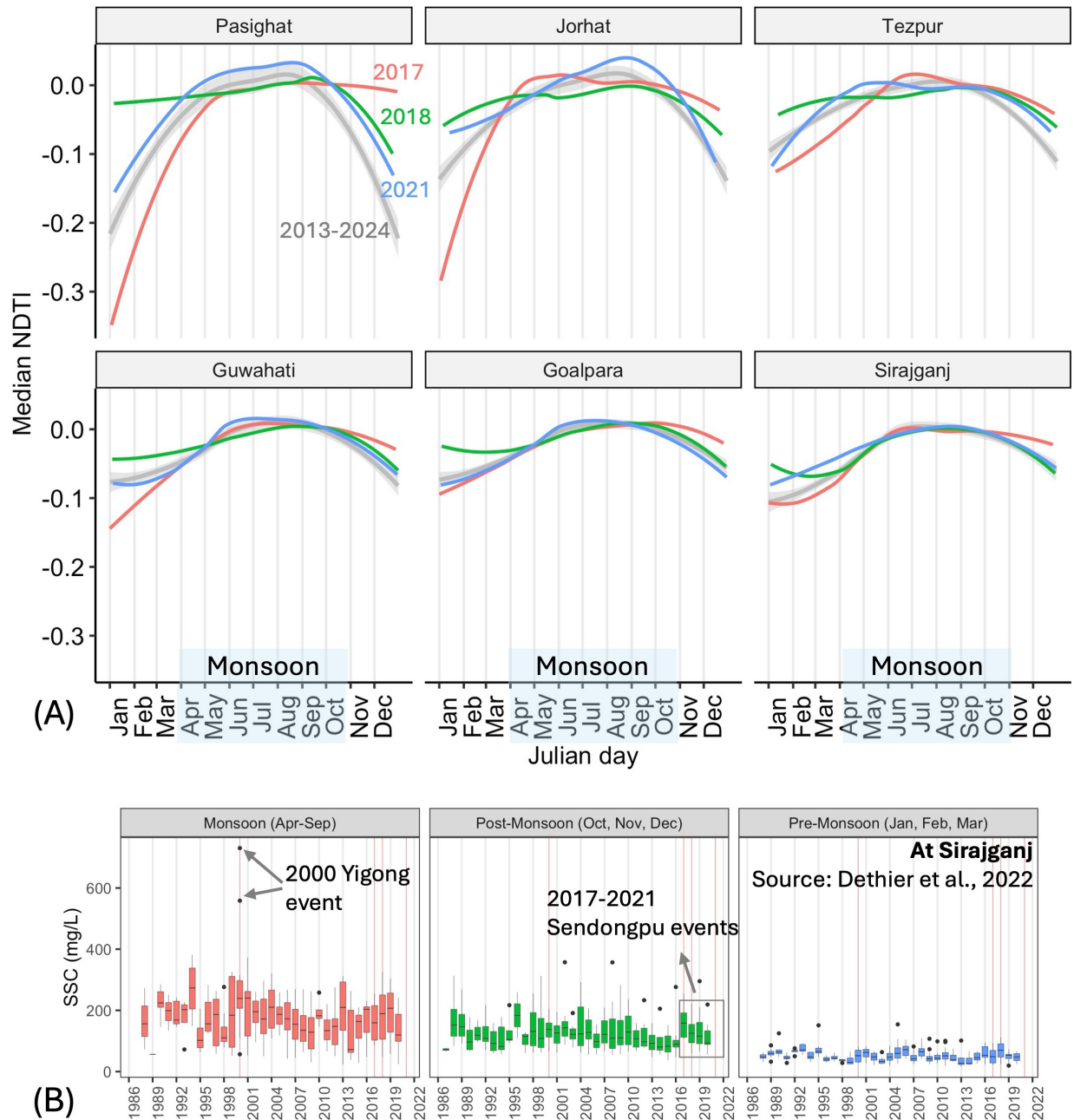


Figure 4: (A) Annual variation of median NDTI, smoothed using the Locally Estimated Scatterplot Smoothing (LOESS) method. Following October 2017, median NDTI exhibited deviations from the 2013–2024 average. This deviation gradually diminished in downstream nodes. Refer to SI Figure S3 for the entire raw time series. (B) Boxplot of suspended sediment concentration at Sirajganj in the delta, separated by seasons (Date source: [Dethier et al. \(2022\)](#)). Boxes represent the interquartile range, while black dots indicate outliers. Red vertical lines indicate years of mass-wasting events.

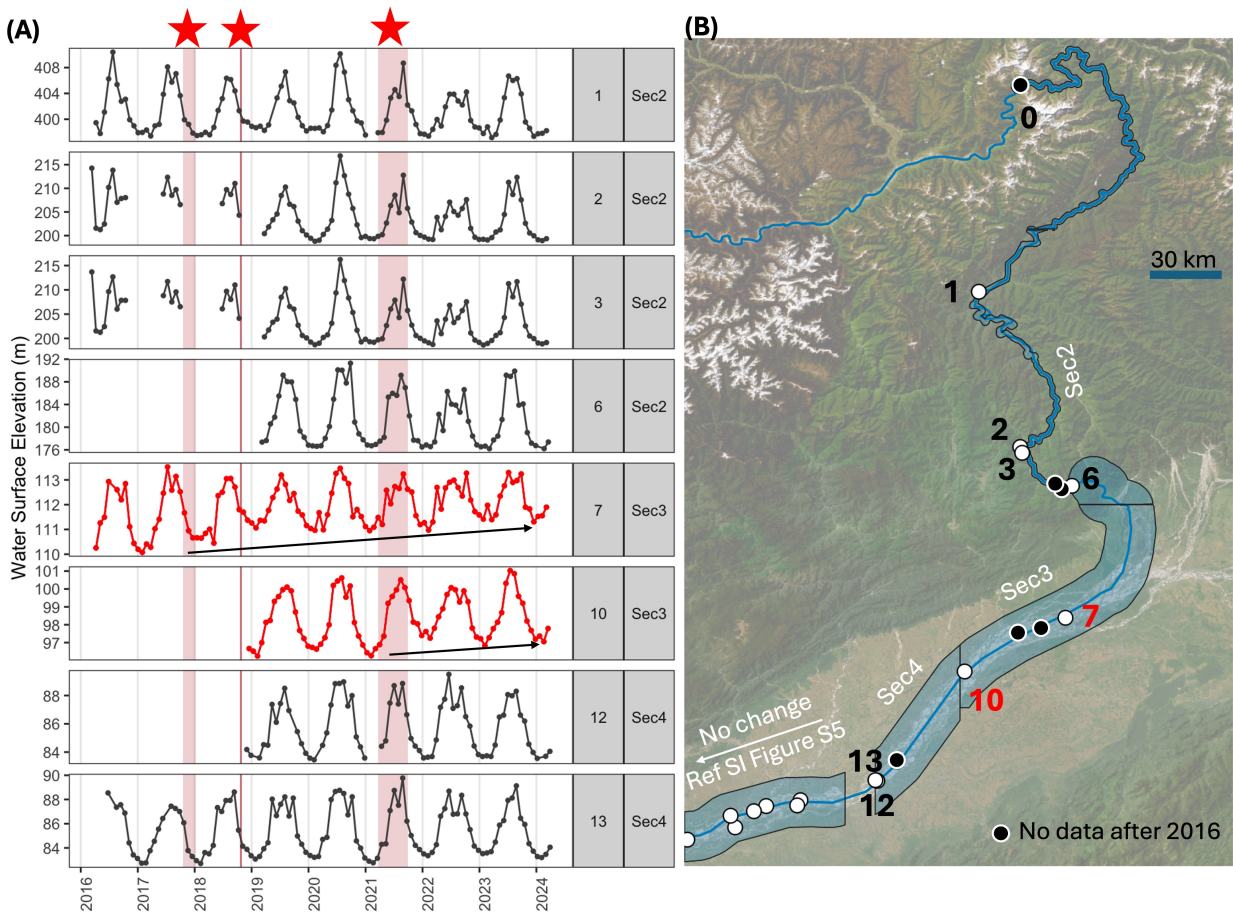


Figure 5: (A) Water surface elevation for selected locations in Sec2, Sec3 and Sec4. Red stars indicate the occurrence of mass-wasting events in the Sendongpu. Lineplots in (A) correspond to locations shown in (B). Only locations, annotated with numbers, have data after 2016, and are presented in lineplot. Refer to SI Figure S5 for entire time series for all locations. Basemap source: Esri, Maxar, Earthstar Geographics, and the GIS User Community.

312 **4.3 % Sandbar area**

313 Figure 6 shows the % Sandbar area variation in Sec1, Sec2 and Sec3. SI Figure S6 shows the time
314 series of all other locations. Consistent with the water surface elevation data, the % Sandbar area
315 shows an increasing trend in Sec3. Following the 2017 event, the percentage of sandbar area in-
316 creased steadily from 20% to 35% during the non-monsoon months from 2018 to 2020 (Figure 6).
317 However, after 2020, the events of 2021 did not result in a further increase in sandbar area. In-
318 stead, the exposed sandbar area decreased from 35% to 25%, followed by a subsequent increase
319 to approximately 30%.

320 **4.4 SAR maximum water extent**

321 As there were observed changes in the Sec3, we investigated the water extent in that section (Figure
322 7). The black pixels in the SAR imagery indicate low backscatter, which is characteristic of water due
323 to its smooth surface (specular reflection) and high dielectric constant. Notably, the water extent in-
324 creased sharply downstream of Pasighat in 2017 (Figure 7), likely triggered by the 2017 Sendongpu
325 mass-wasting event. This trend persisted in subsequent years, with channel width continuing to
326 increase due to ongoing bank and bar erosion. For instance, the eastern channel width expanded
327 from approximately 2.5 km in 2015 to 5 km by 2024. Similarly, the area of an in-channel sandbar
328 decreased significantly, shrinking from 90 km² to 37 km².

329 **5 Discussion**

330 **5.1 Spatial and temporal extent of river response to mass wasting**

331 Mass wasting events generate large quantities of sediment that rivers must either accommodate or
332 transport. Increasing trends of turbidity, water surface elevation and sandbar area are observed in
333 approximately 100 km reach of Sec3 (Figure 8D). Sec3 is located just downstream of the mountain
334 front where the Siang River emerges and transitions into the braided belt of the Brahmaputra (Fig-
335 ures 8A and 3C). Further downstream, in sections like Sec7 (Figure 8C), an increase in turbidity was
336 not coincident with the changes in planform or water surface elevation. Near the delta in Sec10 (Fig-
337 ure 8B), suspended sediment concentrations during post-monsoon months were higher compared
338 to the year preceding the mass-wasting event. These observations suggest that planform changes
339 are primarily limited to the upstream sections near the mountain front, but suspended sediment

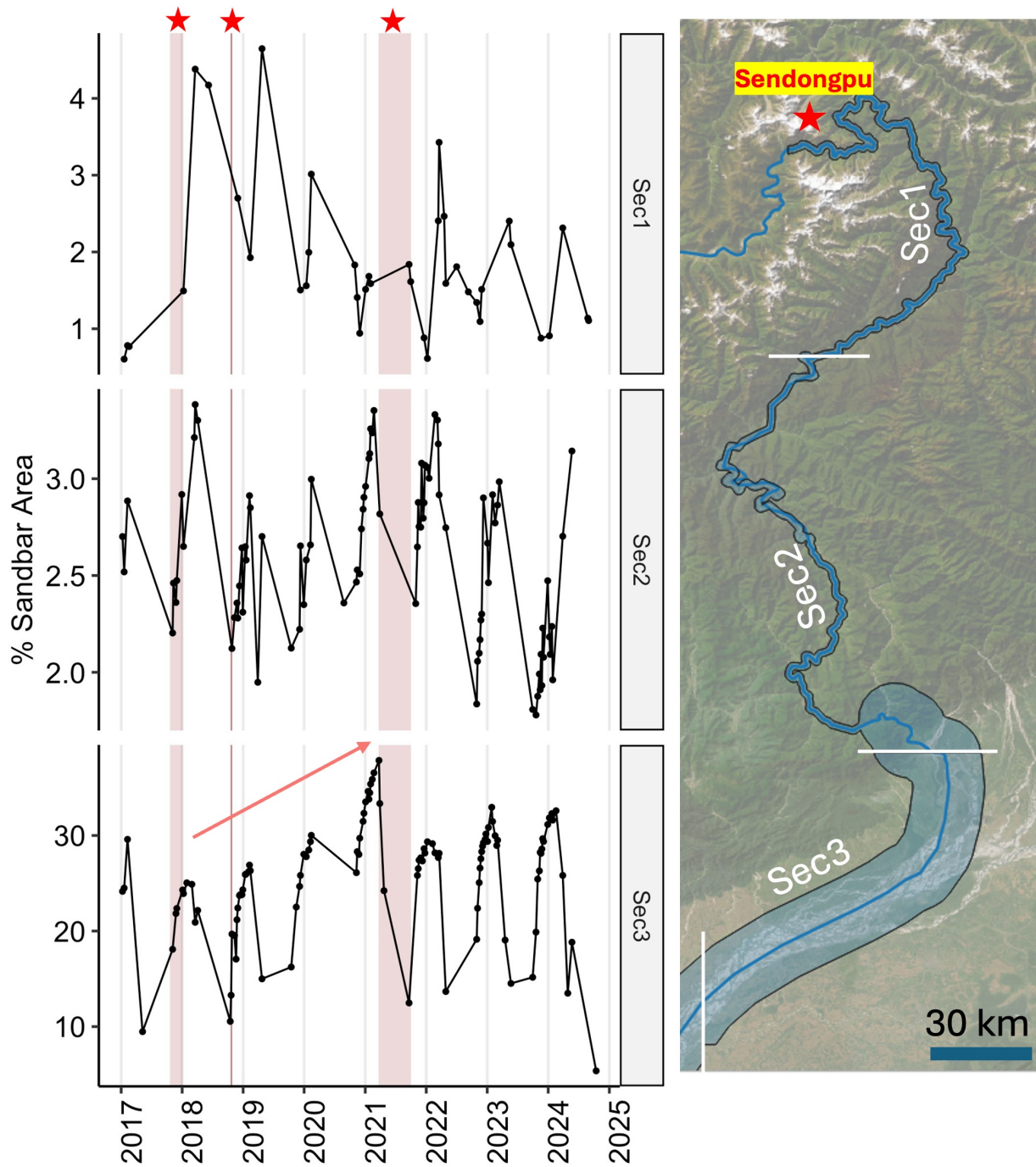


Figure 6: Percentage of sandbar area in Sec1, Sec2 and Sec3. Vertical red area and red star indicate the timing of the mass-wasting events. Basemap source: Esri, Maxar, Earthstar Geographics, and the GIS User Community.

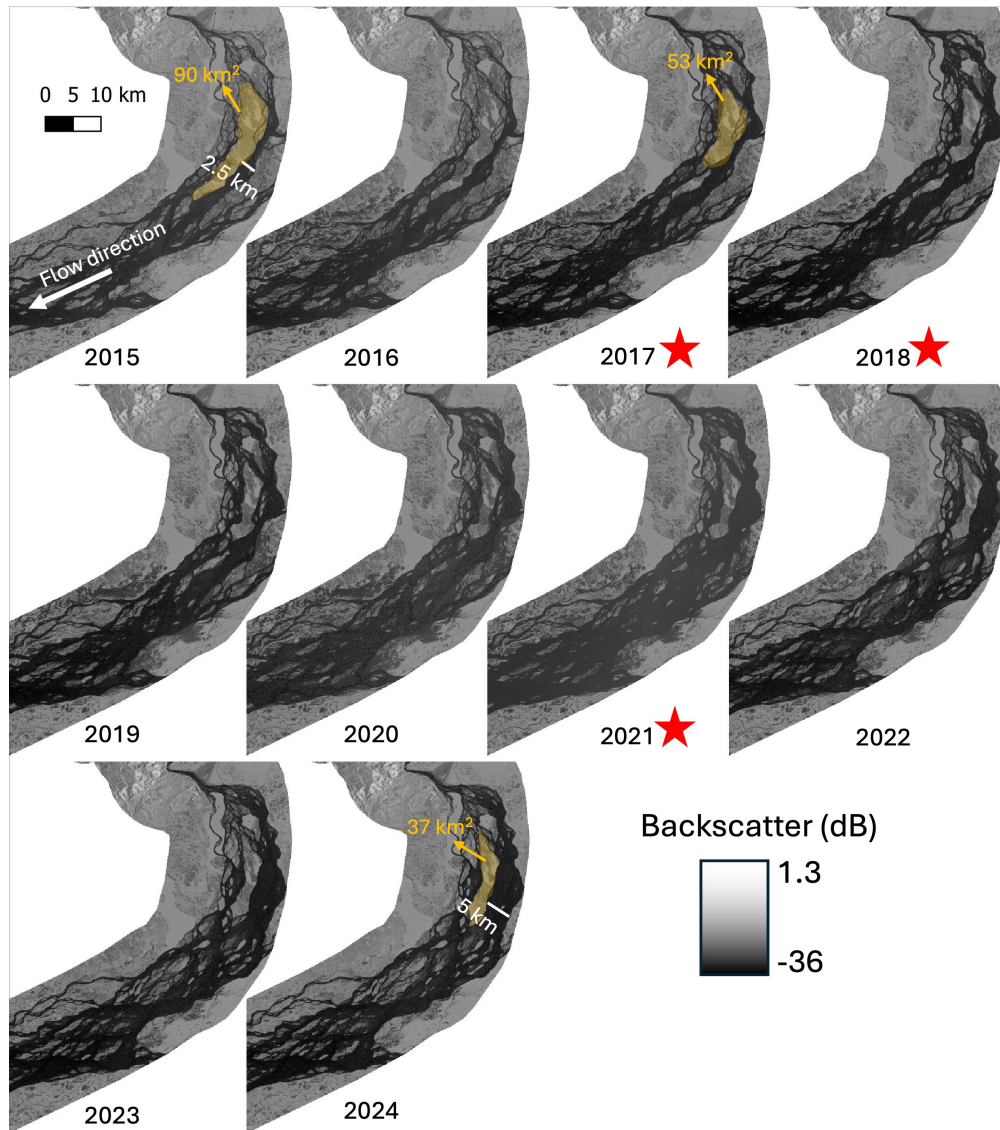


Figure 7: Maximum water extent derived from SAR imagery for each year in the Sec3 region downstream of Pasighat.

340 can be traced far downstream, reaching approximately 1300 km to the delta (Figure 8B). Although
341 attributing downstream signals to specific mass-wasting events in such a large and dynamic fluvial
342 system is inherently challenging, the temporal coincidence between the mass-wasting period and
343 the observed increase in turbidity suggests a strong link. Sediment storage and delayed remobiliza-
344 tion within the river network may also modulate the downstream signal (e.g., ?), particularly when
345 multiple disturbances occur within a short time span.

346 This reason of far-scale transport of the suspended sediments can be linked to the grain size.
347 Suspended sediments in the Brahmaputra River typically range from 10–30 μm (SI Figure S7, Dixit
348 et al., 2023) and are ubiquitous in the water column, regardless of seasonal discharge variations
349 (Garzanti et al., 2011; Lupker et al., 2011). These fine particles may be classified as wash load,
350 which, by definition, can be transported over long distances downstream and can retain signatures of
351 mass-wasting deposits. For instance, based on the sediment provenance analysis, Dixit et al. (2023)
352 suspected the 2021 Sendongpu event's sedimentary signature in suspended sediments sampled at
353 Guwahati.

354 In contrast, the coarse sediments found on Brahmaputra's sandbars typically range from 100–300
355 μm (SI Figure S7, Garzanti et al., 2011; Dixit et al., 2024). These coarser grains are transported as uni-
356 form or graded suspension near the riverbed, primarily during monsoon periods (Dixit et al., 2024).
357 Due to their higher grain size and limited transport efficiency, coarse sediments are predominantly
358 deposited near the mountain front or tributary confluences, where they induce morphological dis-
359 turbances such as channel widening and bar configuration, as observed in Sec3.

360 Nonetheless, the far-scale transport of suspended sediment to the delta is remarkable. Such a
361 phenomenon aligns with efficient sediment evacuation in the Brahmaputra basin reported previ-
362 ously (Singh and France-Lanord, 2002; Garzanti et al., 2004; Finnegan et al., 2008). Notably, the
363 area between Sendongpu and Sec3 brackets the Eastern Syntaxis zone of the Himalayas (Figure
364 8A). This region is characterized by a steep elevation drop from approximately 3000 m to 150 m
365 over a longitudinal distance of 400 km. Such a sharp topographic gradient promotes the efficient
366 transport of sediments through the Himalayan valley (Sec1 and Sec2). These sediments are then
367 deposited at the beginning of the braided belt in Sec3 within a stretch of approximately 100 km,
368 where the river's energy dissipates abruptly due to the sudden change in slope. Provenance anal-
369 yses of the Brahmaputra sands identify Sec3 as a tectonic trap that stores substantial volumes of
370 upstream sediment (Yu et al., 2025).

371 In summary, while fine sediments from a mass-wasting event 1300 km upstream can travel to

372 the delta and elevate turbidity along the entire course (Figure 9), coarse sediments are primarily
373 responsible for morphological disturbances within the first approximately 100 km downstream of
374 the mountain front. Consequently, mass-wasting events not only increase turbidity but also raise
375 water surface elevation and expand sandbar area in this critical reach.

376 **5.2 Implications of mass-wasting for flood risk**

377 Our analysis reveals a significant increase in flood extent downstream of the mountain front (Sec3)
378 following the 2017 event (Figure 7 and SI Figure S8A). Optical imagery from non-monsoon months
379 further highlights an increase in braided-channel width and exposed sandbar area, both triggered by
380 the 2017 Sendongpu mass-wasting events (SI Figure S8B). This increase in water extent cannot be
381 attributed to hydrological factors such as rainfall, as the rainfall distribution over this period shows
382 no notable variability (SI Figure S9). Instead, the observed changes in maximum water extent are
383 clearly linked to the mass-wasting events.

384 Mass-wasting events led to two main types of sediment deposition: vertical channel infilling
385 and lateral sedimentation along bars and banks. Channel infilling likely raised the water surface
386 and increased water extent, while lateral deposition may have contributed to overbank buildup and
387 the formation of alluvial ridges, which can limit flood spread. At the same time, enhanced lateral
388 mobility through bank and bar erosion may also have expanded the water extent. Although the
389 relative influence of these processes is difficult to quantify with the available data, the 2017 mass-
390 wasting event appears to be the key trigger. This event deposited a large volume of sediment in Sec3,
391 as indicated by the increase in sandbar area and water surface elevation. These changes may have
392 altered the sedimentary–hydromorphic regime of Sec3, promoting additional bank and bar erosion,
393 and collectively contributing to the observed increase in maximum water extent. However, by 2021,
394 the sandbar area declined, likely due to the reworking of bars as channels widened and became
395 more interconnected (SI Figure S8B, 2021–2022). This adjustment phase reflects the transition
396 from initial sediment deposition to a more dynamic, laterally mobile system.

397 Interestingly, the eastern channel of the river exhibited more pronounced widening compared
398 to the western channel. Such variability near the mountain front is well-documented in Himalayan
399 rivers (e.g., [Sinha, 2009](#)), and often depends on factors such as longitudinal versus across-channel
400 slope and the relative elevation of both channels ([Gearon et al., 2024](#)). Although high-resolution
401 elevation data are unavailable to confirm these factors definitively, our observations align with long-
402 term trends. For example, [Devi et al. \(2016\)](#) documented an increase in maximum channel width

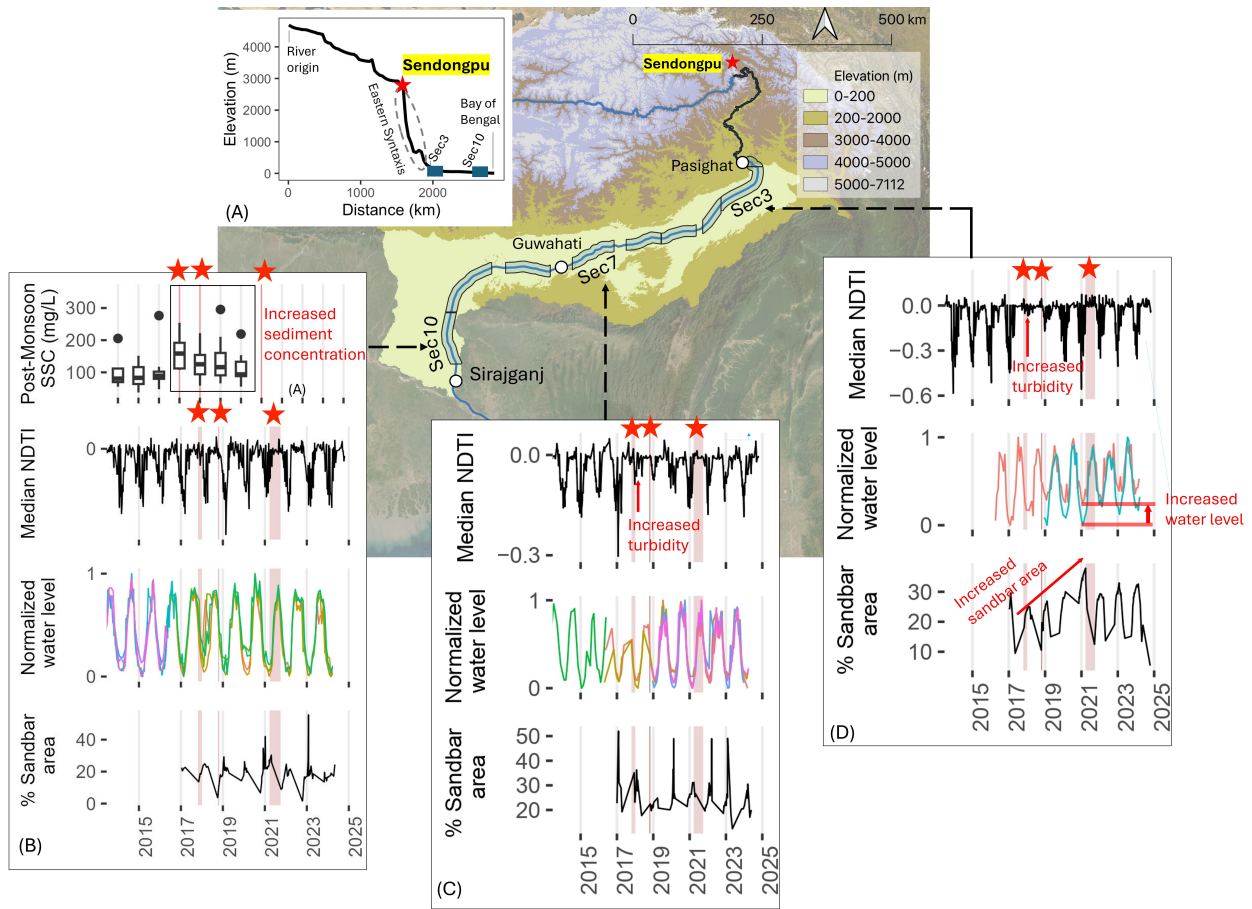


Figure 8: Map depicting elevation values across the Brahmaputra basin. (A) Longitudinal profile of the Yarlung-Siang-Brahmaputra River, with key locations highlighted along the profile. The region between Sendongpu and Pasighat marks the Eastern Syntaxis, characterized by a sharp elevation gradient. (B), (C), and (D) illustrate the variation in key parameters—such as turbidity, suspended sediment concentration and water surface elevation—in response to mass-wasting events at Sec10, Sec7 and Sec3, respectively. Different colors in the normalized water level plot indicate different locations in their respective sections. Basemap source: Esri, Maxar, Earthstar Geographics, and the GIS User Community.

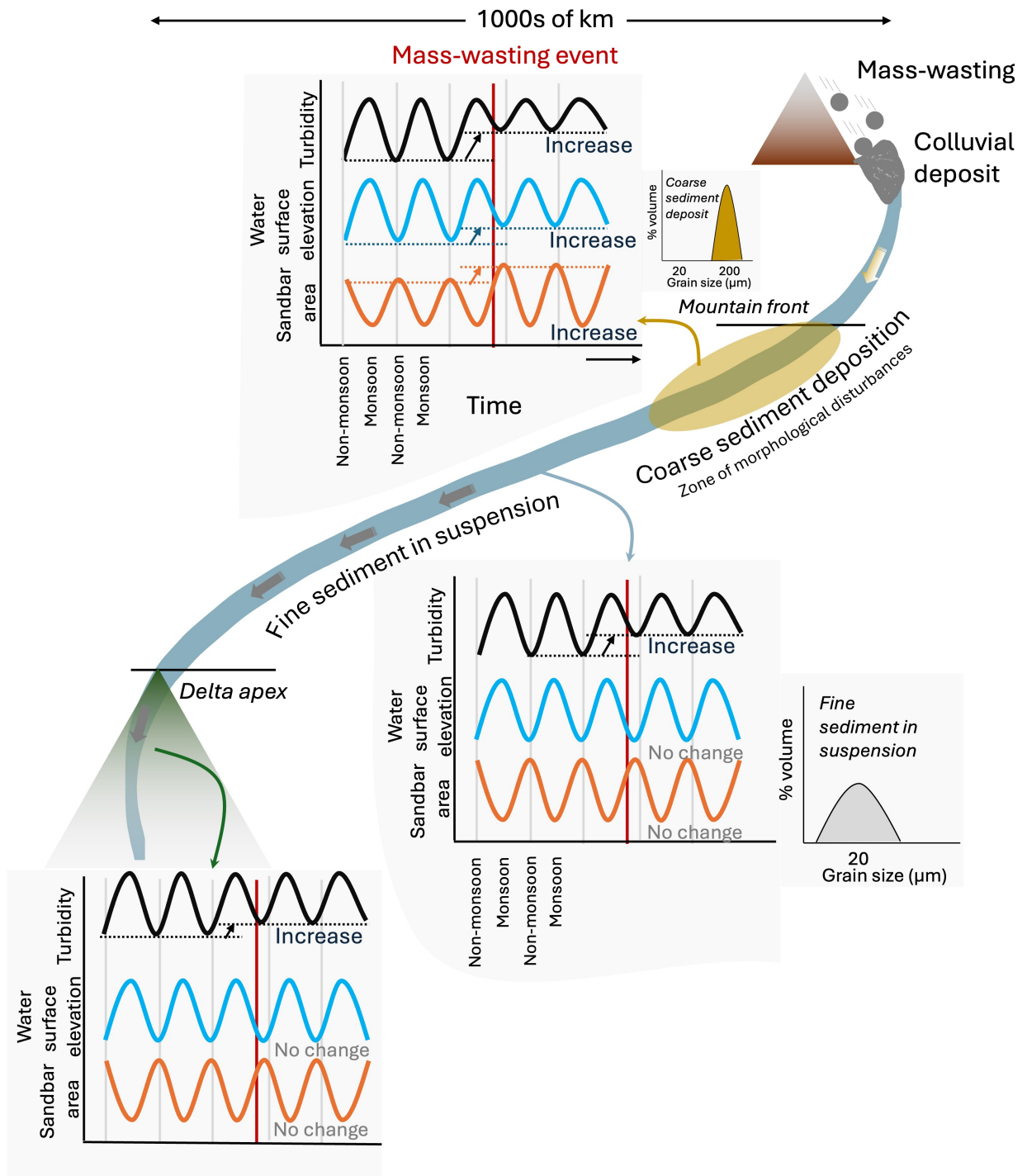


Figure 9: Conceptual figure illustrating the key findings. Near the mountain front, coarse sediments from mass-wasting events generate morphological disturbances, raising water surface elevations and increasing sandbar area. In contrast, fine sediments are transported across the basin and reach the delta, even in a large river system such as the Brahmaputra.

Table 2: Comparison of Sendongpu (2018) and Yigong (2000) mass-wasting related events. Data source. Data sources are [Guo et al. \(2023\)](#); [Li et al. \(2024\)](#); [Chen et al. \(2020\)](#).

	Sendongpu event 2018	Yigong event 2000
Lake volume	$490 \times 10^6 \text{ m}^3$	$3000 \times 10^6 \text{ m}^3$
Colluvial deposit volume	$35 \times 10^6 \text{ m}^3$	$300 \times 10^6 \text{ m}^3$
Breakage	Partial breakage, river maintained flow	Sudden breakage, outburst flood
Response	Increased turbidity in subsequent post-monsoon	High turbidity during monsoon of 2000

403 from 1 km to 2.7 km between 1966–1972 and 2012 in the Sec3. [Yu et al. \(2025\)](#) report the Sec3 region as the tectonic trap which stores upstream sediments. Therefore, such morphological changes
 404 are more likely to occur.
 405

406 **5.3 Comparison of Yigong and Sendongpu events**

407 The differing impacts of the Yigong and Sendongpu events on river morphology and sediment dynamics are notable and can provide key understanding of how rivers respond to large-scale mass-wasting events. The characteristics of these events and responses are summarized in Table 2. The
 408 Yigong event of 2000 was a significantly larger outburst event, characterized by the breaching of a
 409 $3000 \times 10^6 \text{ m}^3$ lake and a $300 \times 10^6 \text{ m}^3$ landslide dam. In contrast, the Sendongpu events, such as
 410 the 2018 outburst, involved a $490 \times 10^6 \text{ m}^3$ lake and a much smaller debris dam of approximately 35
 411 $\times 10^6 \text{ m}^3$ ([Chen et al., 2020](#); [Li et al., 2024](#)). The subsequent outburst of the Yigong event caused a
 412 sudden influx of sediments, leading to sediment concentrations three times higher than usual downstream, reaching as far as approximately 1300 km to the delta (Figure 4B). In contrast, the rates of
 413 sediment mobilization from Sendongpu events were relatively slower and lacked a sudden outburst,
 414 resulting in a more subdued and gradual sediment response.
 415
 416
 417

418 **6 Conclusion**

419 Investigation of river response to mass-wasting events is often limited at reach- or catchment-scale.
 420 In this study, we investigated the river response of a large river to major mass-wasting events in
 421 the Eastern Himalaya over extended spatial (approximately 1000 of kilometer) and temporal (years)
 422 scales, leveraging an advanced cloud computing platform, Google Earth Engine. We examined how
 423 these events leave signatures in suspended sediment dynamics, water surface elevation, river sand-

424 bar morphology and flood inundation.

425 The Eastern Syntaxis in the Himalaya is a region of unique and large-scale mass-wasting pro-
426 cesses both in modern and ancient time scales. Over the past seven years, the Sendongpu glacier
427 alone has eroded approximately 50, 180 and $329 \times 10^6 \text{ m}^3$ of sediment during three major events
428 in 2017, 2018 and 2021, respectively. Additionally, the nearby Yigong event eroded $300 \times 10^6 \text{ m}^3$
429 in 2000 following a landslide outburst flood.

430 Our findings demonstrate that suspended sediment signatures from mass-wasting events can be
431 traced up to approximately 1000 km downstream, reaching the delta. However, coarse sediments
432 primarily influence river planform changes in the approximately 100 km reach near the mountain
433 front, where the river starts to establish a braided belt. In this region, we observed an increase
434 in sandbar area of between 20% and 35% following the 2017 events. This morphological change
435 coincided with elevated water surface elevation and high flood inundation, suggesting significant
436 channel infilling.

437 Comparing the Sendongpu and Yigong events highlights how the character of an individual event
438 can define the river's response. For example, the 2000 Yigong event, a single large and sudden out-
439 burst, led to a far-reaching but short-lived sediment response. In contrast, the Sendongpu events,
440 which were smaller, slower and more frequent, produced a prolonged but less intense river response,
441 with sediment retained in valleys and gradually mobilized over time.

442 While we quantified the spatial extent of river responses to mass-wasting events, future studies
443 should focus on quantifying sediment budgets specific to these events. This includes determining
444 the volume of sediment generated by mass wasting and assessing its partitioning among storage in
445 valleys, floodplains and river channels. Such data are critical for understanding the impacts of mass-
446 wasting events on flood management or hydropower planning in regions that are already flood-
447 prone and earmarked for extensive hydropower development. These changes are likely to reshape
448 river systems in ways that we have not yet been able to fully observe or predict.

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