

# Geophysical Research Letters®











## RESEARCH LETTER

10.1029/2022GL102684

## River Delta Morphotypes Emerge From Multiscale Characterization of Shorelines

### Key Points:

- Three process-informed, geometric, and spectral metrics are introduced to characterize multiscale shoreline features
- Unsupervised clustering of the shoreline metrics reveals five distinct delta morphotypes which correspond to distinct forcings
- Morphotypes can be robustly estimated from readily available satellite imagery

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### Supporting Information:

Supporting Information may be found in the online version of this article.

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### Citation:

Vulis, L., Tejedor, A., Ma, H., Nienhuis, J. H., Broaddus, C. M., Brown, J., et al. (2023). River delta morphotypes emerge from multiscale characterization of shorelines. *Geophysical Research Letters*, 50, e2022GL102684. <https://doi.org/10.1029/2022GL102684>

Received 4 JAN 2023  
Accepted 28 FEB 2023

**Abstract** Delta shoreline structure has long been hypothesized to encode information on the relative influence of fluvial, wave, and tidal processes on delta formation and evolution. We introduce here a novel multiscale characterization of shorelines by defining three process-informed morphological metrics. We show that this characterization yields self-emerging classes of morphologically similar deltas, that is, delta morphotypes, and also predicts the dominant forcing of each morphotype. Then we show that the dominant forcings inferred from shoreline structure generally align with those estimated via relative sediment fluxes, while positing that misalignments arise from spatiotemporal heterogeneity in deltaic sediment fluxes not captured in their estimates. The proposed framework for shoreline characterization advances our quantitative understanding of how shoreline features reflect delta forcings, and may aid in deciphering paleoclimate from images of ancient deposits and projecting delta morphologic response to changes in sediment fluxes.

**Plain Language Summary** It has long been hypothesized that delta morphology and its evolution are primarily dictated by the interplay of three primary forcings: riverine, wave, and tidal. The rich diversity of delta morphologies observed across the globe is attributed to the relative dominance of those forcings. This study aims to retrieve the forcing information archived in delta morphology, specifically in the delta's shoreline, to establish a morphologic classification of deltas. To accomplish that, we propose a set of morphologic metrics targeting the specific locations and scales of the shoreline reworked by riverine, wave, and tidal processes. We show that the projection of the complex shoreline geometries into the low-dimensional space defined by the proposed morphologic metrics leads to clusters of deltas with similar features, denoted here as morphotypes. Guided by the values of the metrics, we can infer the dominant forcings underlying each morphotype, which generally align with those obtained by estimating the sediment fluxes transported by each forcing. Misalignments are also observed and are mostly due to space-time heterogeneities and uncertainties in the quantified sediment fluxes. Our methodology relies on readily-available remote sensing images and is shown to have low sensitivity to its defining parameters, offering insight into predicting deltaic geomorphic response to changing forcings.

## 1. Introduction

River deltas are complex ecogeomorphic landscapes which are home to upwards of 300 million people due to their fertile soils and rich ecosystems (Edmonds et al., 2020). Their intricate hydromorphology controls near-shore biogeochemical function (Knights et al., 2020; Zoccarato et al., 2019), connectivity between surface and subsurface hydrogeology and reservoirs (Sawyer et al., 2015), coastal resilience (Hoitink et al., 2020; Tognin et al., 2021), and ecosystem services (Adams et al., 2018). Deltas are particularly vulnerable to climate change due to their low relief, coastal proximity, and large populations (Edmonds et al., 2020; Hoitink et al., 2020). It is therefore critical to understand how sea level rise and changing riverine sediment loads will impact these systems (Chadwick et al., 2020; Nienhuis et al., 2023) and toward this goal, developing a quantitative framework that links the driving forces forming deltas to delta morphology and function is imperative.

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Fifty years ago, Galloway introduced the paradigm that river deltas are shaped by the interplay of progradational riverine forcings and erosional marine (wave and tide) forcings, which has steered subsequent research on river delta evolution (Galloway, 1975; see also e.g., Ainsworth et al., 2011; Anthony, 2015; Bhattacharya & Giosan, 2003; Nienhuis et al., 2020; Seybold et al., 2007; Syvitski & Saito, 2007). The relative balance of these forcings and the multiple spatio-temporal scales at which they operate result in a stunning degree of variability in shoreline structure and channel network geometry and topology (Fagherazzi et al., 2015; Hoitink et al., 2017; Jerolmack & Swenson, 2007; Konkol et al., 2022; Tejedor et al., 2016, 2017). Rivers act to prograde the delta platform at large scales and increase roughness at fine scales via the growth of mouth bars and distributary channel expansion (Fagherazzi et al., 2015; Wolinsky et al., 2010). Waves generate alongshore transport that diffuse sediment along the shoreline at fine scales but can lead to spits at coarser scales (Ashton & Giosan, 2011) and suppress mouth-bar development (Jerolmack & Swenson, 2007). Tidal forces widen distributary channels and construct headless channels that lack connections to the upstream river, roughening the shoreline at multiple scales (Hoitink et al., 2017; Nienhuis et al., 2018).

Recently, the relative magnitudes of the forcings in the Galloway framework have been quantified via a sediment flux approach (Nienhuis et al., 2020). However, shoreline shape, a crucial ingredient in the qualitative morphological classification originally posed by Galloway (see Table 2 in Galloway, 1975), has not been quantified in a way to differentiate between visually distinct deltas, nor has been shown to have a clear relationship with forcings (e.g., Baumgardner, 2016). This is in part because analysis of shoreline structure has typically focused on a single length scale using metrics such as shoreline variability (Straub et al., 2015) roughness or rugosity measures (Baumgardner, 2016; Caldwell & Edmonds, 2014; Geleynse et al., 2012), and shape factors (Lauzon et al., 2019; Nienhuis et al., 2015; Wolinsky et al., 2010). Such metrics do not necessarily measure shoreline structure at process length scales, nor do they capture the multiscale variability caused by the interplay of the three driving forces.

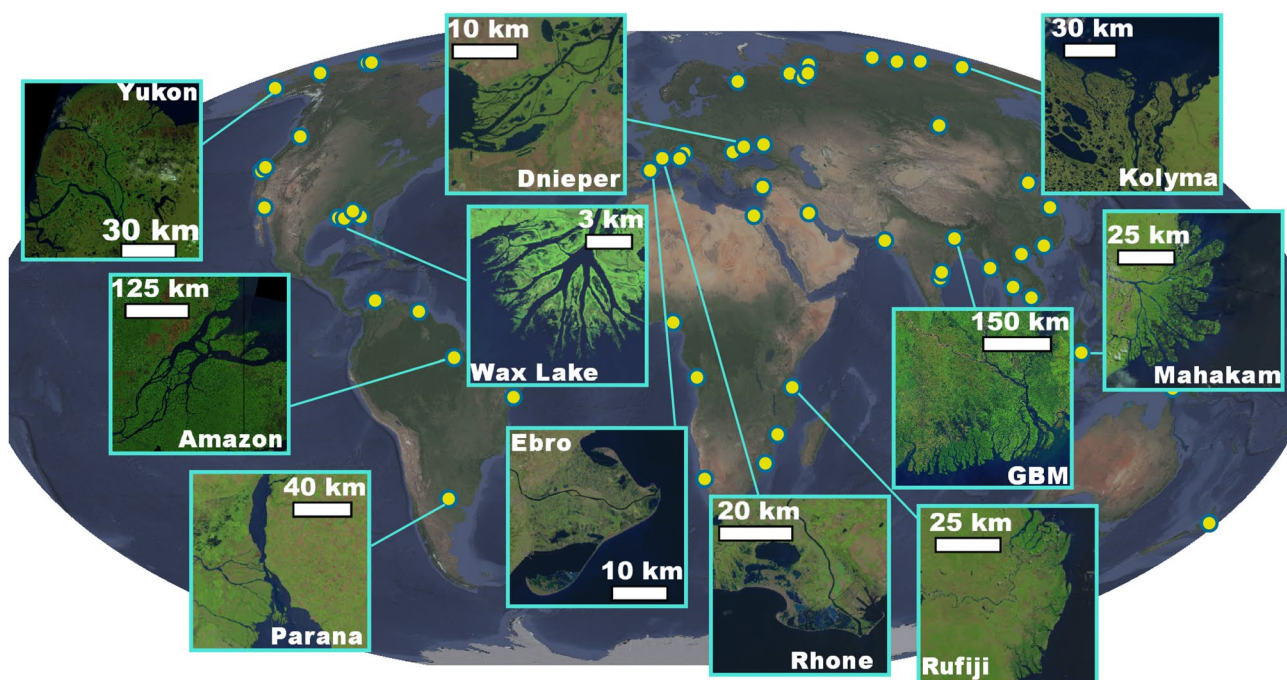
Here, we propose a set of process-informed, multiscale metrics of river delta shoreline shape that combine geometric and spectral measures to develop a quantitative classification of delta morphology. Our approach utilizes localized analysis of shoreline structure both in the space and wavenumber domains to isolate features corresponding to different processes acting at multiple scales. Unsupervised clustering of the shoreline morphometrics identifies five classes of morphologically similar deltas, that is, delta morphotypes. Based on the values of the process-informed metrics, dominant forcings are attributed to each morphotype, which we then show to generally align with the dominant forcings quantitatively estimated by their relative sediment fluxes (see Text S1 in Supporting Information S1 for details; Nienhuis et al., 2020). We hypothesize that misalignments between the two are due to spatiotemporal heterogeneity in the sediment fluxes which are not captured by their estimated values.

## 2. Multiscale Characterization of Delta Shorelines

We analyzed the shorelines of 54 global river deltas sampled across a representative range of sizes and morphologic variability, including systems with wave-influenced, river-influenced, and tide-influenced features (see Table S1; Figure 1; Syvitski & Saito, 2007). River delta shorelines were defined using the Opening Angle Method (OAM) with a critical angle of  $45^\circ$  (Shaw et al., 2008). To define a shoreline, the OAM requires a binary water mask, which was obtained by thresholding water occurrence masks from the Landsat-derived, 30-m spatial resolution Global Surface Water data set (Pekel et al., 2016).

We defined three scales at which delta shoreline structure exhibits variability, which are linked to the balance of river, tide, and wave forcings: a macroscale (overall delta platform), mesoscale (mouth width scale), and microscale (beach scale). We developed metrics to capture the variability at those scales as discussed below.

At the macroscale, riverine sediment deposition leads to delta progradation and growth into the receiving basin and generates extrusional shapes (i.e., convex shoreline; Caldwell & Edmonds, 2014; Galloway, 1975). When wave-driven alongshore transport removes the majority of riverine sediment flux, the delta has no protrusion, and is linear (i.e., mostly flat shoreline; Nienhuis et al., 2015). Lastly, tidal forcings erode subaerial sediment into the nearshore and construct a subaqueous platform (Hoitink et al., 2017). This net erosion from land leads to a funnel-shaped, concave subaerial delta, or estuary, which intrudes into the surrounding landscape (i.e., a concave shoreline). We therefore measured the curvature of the entire shoreline (Figure 2; Jammalamadaka &



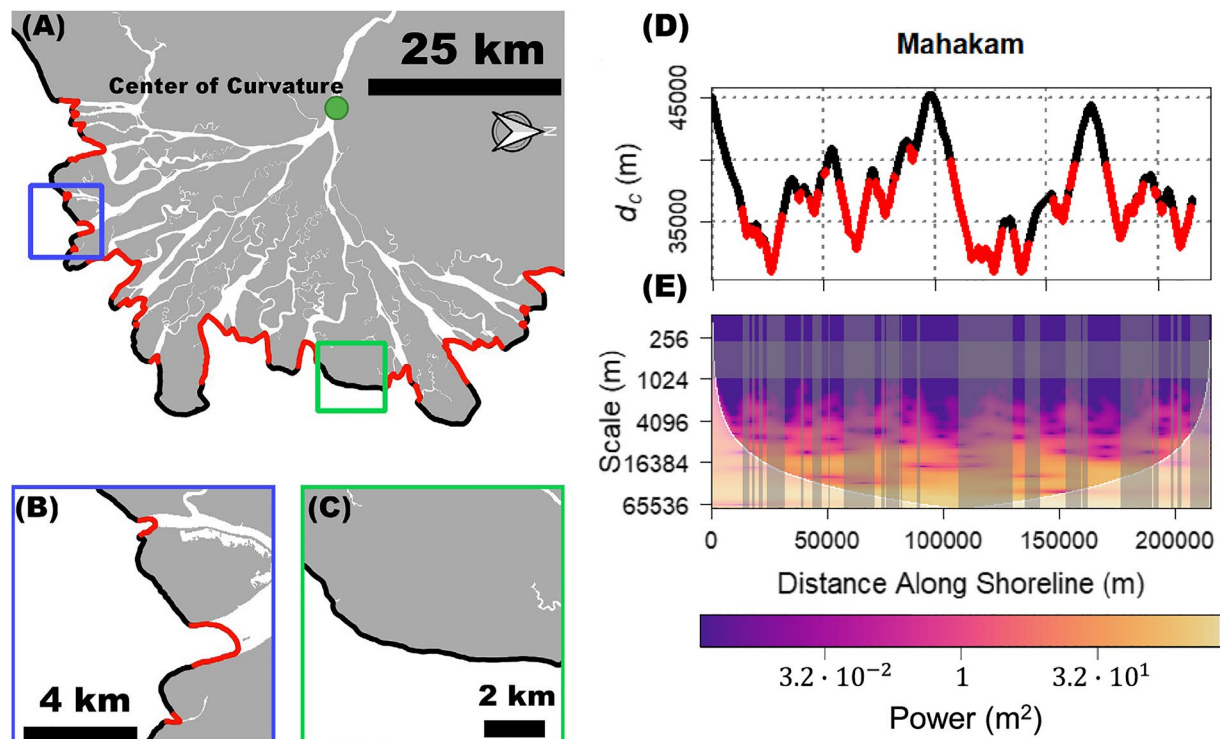
**Figure 1.** The morphologic variability of Earth's deltas. River deltas show differences in shoreline structure attributed to the relative balance of river, wave, and tidal sediment fluxes (Galloway, 1975). Yellow dots show locations of a globally distributed sample of 54 deltas analyzed in this study. Satellite imagery courtesy of Landsat and Google Earth.

SenGupta, 2001), to classify deltas as convex (extrusional), concave (intrusional), or flat (see Text S1 in the Supporting Information S1).

At the mesoscale, the influence of rivers, waves, and tides on channel mouths dictates multiple intermediate scales of variability on the shoreline. Tidal forces widen mouths exponentially (Nienhuis et al., 2018) which leads to multiscale undulations in the shoreline (e.g., Amazon and Ganges-Brahmaputra, GBM, deltas; Figure 1). Rivers form mouth bars and bifurcations leading to small but numerous mouths, which result in intermediate to fine scale undulations in the shoreline (e.g., Dnieper delta). Lastly, wave-driven sediment transport prevents mouth bar formation (Jerolmack & Swenson, 2007) and reduces the number of channels (Broaddus et al., 2022), resulting in long shorelines with few, small undulations (e.g., Ebro delta). To measure the contribution of mouths to the overall variability of the shoreline structure, we first projected the shoreline into a univariate spatial-series by recording the distance from each point along the shoreline to the center of curvature of the macroscale shape of the delta (Figure 2a). Then, we identified sections of the shoreline spatial-series corresponding to the mouths and measured via localized wavelet transforms (Kumar & Foufoula-Georgiou, 1994) the fraction of variance contributed by the mouths,  $fM$  (see Text S1 in Supporting Information S1; Figures 2d and 2e).

Finally, at the microscale, wave-driven alongshore transport diffuses sediment along the coast and smooths shorelines (Ashton et al., 2001), while rivers and tides introduce variability from distributary and headless channels (Wolinsky et al., 2010). Therefore, we measured the fine scale variance ( $FSV$ ), as the variance at wavelengths of 300–1,000 m, to capture these differences (Figure 2e). The lower bound is the result of the minimum reliable scale above which discretization, aliasing, and smoothing effects do not affect the spectra, derived from 30-m spatial resolution Landsat imagery. The upper bound is an approximation of the range of scales within which waves act to smooth shorelines and below which large scale features such as spits begin to emerge. The results are robust to shifting the upper bound from 800 to 1,100 m (Figure S1 in Supporting Information S1). Furthermore, to separate shorelines that may have equal  $FSV$  but relatively more power at larger wavelengths compared with shorelines that have relatively less power at those wavelengths, the  $FSV$  is adjusted by the degree of heterogeneity over the spectral range by multiplying by a spectral Gini coefficient,  $g$ , defining the  $gFSV$ . The spectral Gini coefficient is a measure of the deviation of the spectra from white noise, that is, a random signal with a flat spectrum (Figure S2 and Text S2 in Supporting Information S1). With these three metrics we quantitatively compare the shoreline morphology of river deltas and explore the possible emergence of distinct morphotypes.



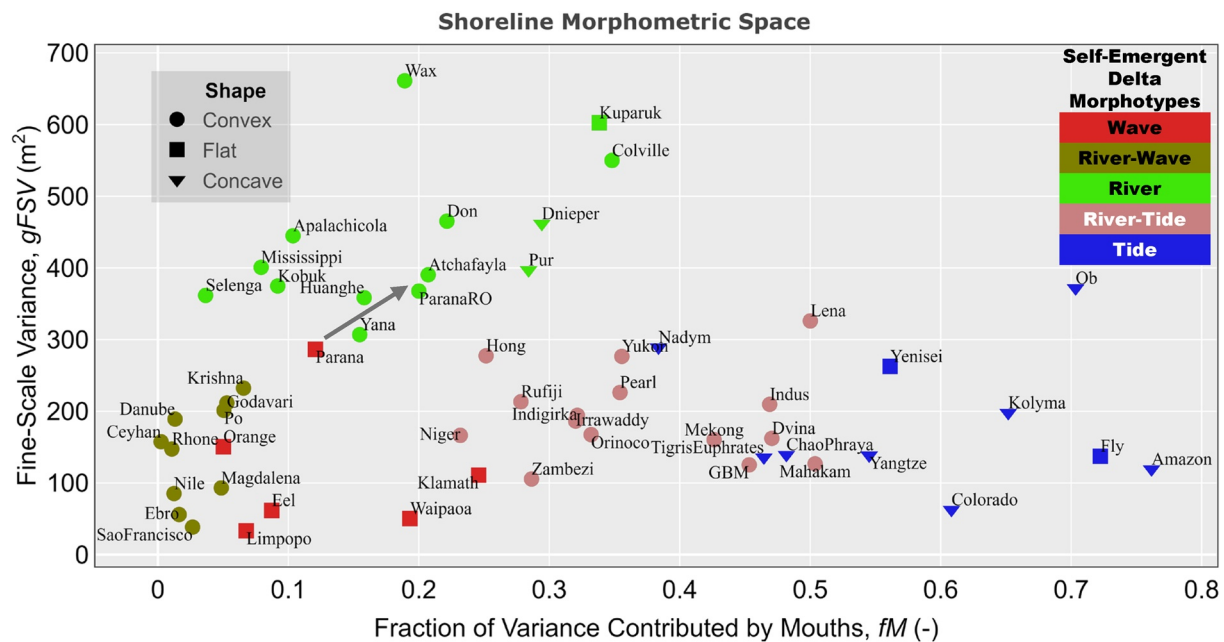


**Figure 2.** Example of the multiscale features of shoreline structure on the Mahakam Delta, Indonesia. (a) The shoreline of the delta, defined using the Opening Angle Method (OAM) with a critical angle of  $45^\circ$ , shows multiple scales of variability. At the macroscale, a delta may be convex due to river deposition, flat due to wave-driven along shore transport, or concave due to tidal widening and estuarine conditions. This is measured here by the ratio between the radius of curvature and the length of the shoreline. (b) Mouths formed by rivers and tides lead to undulations in the shoreline at a scale determined by the relative river and tide fluxes. (c) At the microscale, waves diffuse sediment parallel to the coast and smooth the shoreline, while rivers and tides roughen it. (d) To measure meso- and microscale variability, the 2D shorelines are mapped to a univariate signal defined as the distance,  $d_c(s)$  from each point along the shoreline to the center of curvature, where  $s$  is the distance along the shoreline. (e) The wavelet transform is used to estimate the fraction of variance contributed by the mouths,  $fM$ , which are marked in red in the preceding panels, and the Gini-corrected Finescale Variance  $gFSV$ , that is, the variance from scales (wavelengths) between 300 and 1,000 m.

### 3. Shoreline Morphometric Space

The proposed shoreline metrics construct a three-dimensional Shoreline Morphometric Space (SMS) within which deltas can be positioned and compared (Figure 3). To objectively and robustly identify clusters that categorically classify deltas within this space, we used an unsupervised machine learning algorithm, k-prototypes (a modification of k-means clustering that accounts for categorical predictors such as the macroscale shape; Huang, 1998). Five morphotypes, that is, clusters of morphologically similar deltas, emerge from the three-dimensional SMS (Figure 3) and are displayed in Figure 4. Note that the identified clusters are robust and stable as found by randomly resampling 80% of the deltas and repeating the clustering procedure, obtaining an identical grouping of delta morphologies in 97% of simulations.

The first morphotype is denoted as the “tidal morphotype” as these deltas are concave and flat with mouth-dominated shorelines and low finescale variance, indicative of tide-domination (Figure 4), for example, the Fly and Amazon deltas. It also includes valley-confined deltas like the Ob and Yenisei due to their wide mouths (Figure 4). The second morphotype is denoted as the “river morphotype” as these deltas are characterized by an intermediate fraction of variance contributed by mouths, are rough at fine scales, and have a convex planform, for example, the Selenga and Mississippi deltas (Figure 4). Valley-confined deltas such as the Dnieper and Don, which are concave and flat but have high fine scale variability, are also included as part of the river morphotype. The third morphotype is denoted as the “wave morphotype” as these deltas are flat, lack a subaerial protrusion formed by river deposition, and smooth at fine scales, for example, the Eel and Orange deltas (Figure 4). The fourth morphotype is denoted as the “river-wave morphotype” as these deltas are convex, smooth at fine scales, typically have spits or flying spits, and little to no variability contributed by mouths, for example, the Ebro and Rhone deltas. Lastly, the fifth morphotype is denoted as the “river-tide morphotype” as it contains convex deltas with tidally



**Figure 3.** The Shoreline Morphometric Space (SMS). Deltas shorelines are positioned in the three-dimensional space constructed by the macroscale shape,  $fM$ , and  $gFSV$  metrics. Unsupervised clustering of the SMS using k-prototypes reveals five self-emergent delta morphotypes, that is, classes of morphologically similar systems. The relative position of the deltas in the SMS elucidates the dominant forcing acting on each morphotype, for example, increased  $fM$  a signature of greater tidal influence and low  $gFSV$  a signature of wave influence. The classified deltas are shown in Figure 4. The arrow indicates the shift in the SMS position of the river distributary section of the Parana shoreline (ParanaRO) compared with the shoreline of the entire Parana, see text for details.

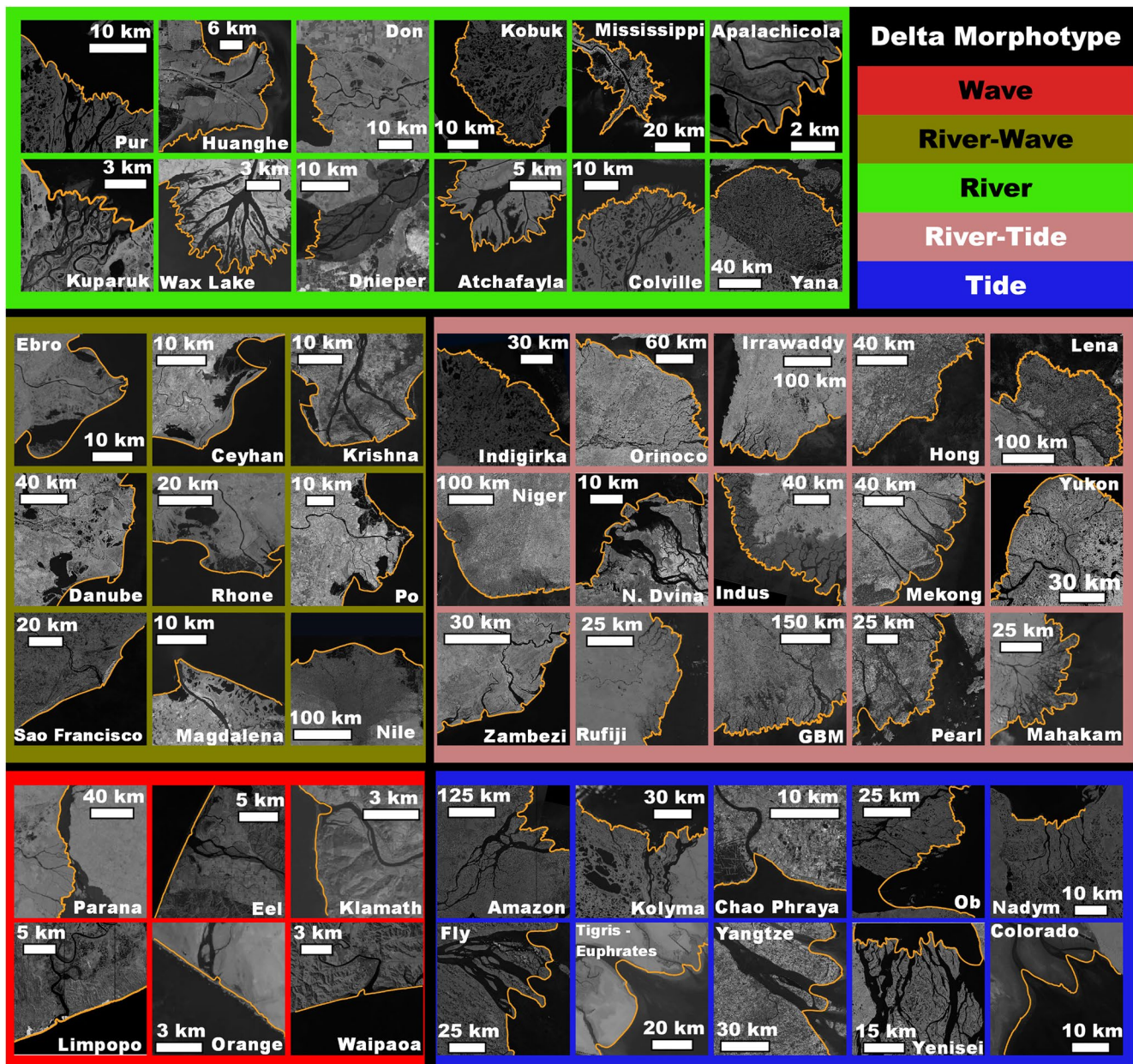
widened mouths and headless channels, resulting in intermediate variability contributed by mouths, for example, the Mahakam and Orinoco deltas.

The dominant forcings determined by the quantitative classification of shorelines correspond with expert assessment of the dominant forcings based on qualitative comparisons of delta morphology (Ainsworth et al., 2011; Nienhuis et al., 2020) suggesting that shoreline structure carries a distinct signature of the processes that generated that delta. An interesting further step is to check whether the inferred dominant forcings align with the relative sediment fluxes driven by each forcing, for which we use the recently developed sediment flux estimation framework of Nienhuis et al. (2020).

#### 4. Are Delta Morphotypes Aligned With Relative Sediment Fluxes?

Each of the 54 deltas was projected onto the ternary Galloway diagram according to the relative sediment flux transported by rivers, waves, and tides as estimated in Nienhuis et al. (2020) (Figure 5). Before contrasting delta morphotypes with their relative sediment fluxes we note a few important issues which we anticipate to cause discrepancies in the mapping between the morphotype and dominant sediment flux. First, the marine sediment fluxes are estimated using simplified, although nonlinear, physical models that transform tidal amplitudes and offshore wave-climate into tidal and wave sediment fluxes, respectively. Therefore, any uncertainty in the tidal amplitude and wave climate will propagate into uncertainty in the sediment flux estimate. Second, sediment fluxes are estimated using single, representative locations for wave climate, tidal amplitude, and fluvial discharge, not acknowledging possible multi-mouth or multi-lobe structure (Nienhuis et al., 2020). Moreover, the sediment fluxes are estimated using contemporary wave climate, tidal amplitude measurements, and modeled, pre-anthropogenically-influenced riverine discharge and sediment loads (Text S1 in Supporting Information S1; Nienhuis et al., 2020), and represent snapshots of the relative sediment flux, while delta morphology represents the temporally integrated effect of the forcings acting on the delta (Syvitski et al., 2022). Accordingly, any significant spatiotemporal heterogeneity or non-stationarity in the fluxes over each delta's evolution might not be reflected in the contemporary sediment flux estimates. Therefore, some misalignments between delta morphotype and dominant sediment flux are expected, hoping however, that a general agreement will emerge.

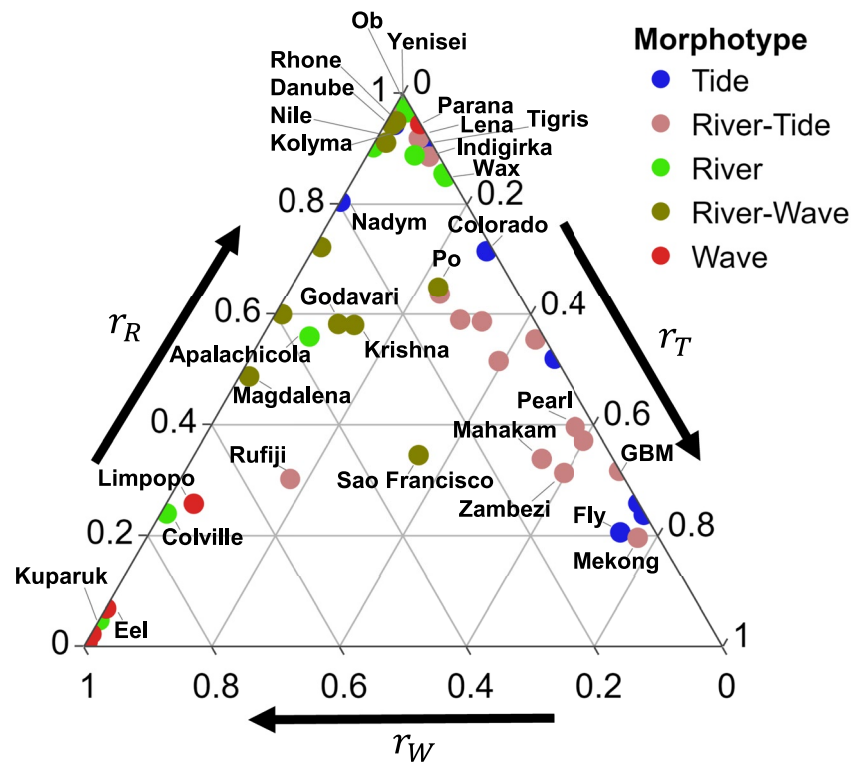




**Figure 4.** Deltaic morphotypes identified from the Shoreline Morphometric Space (SMS). The deltas corresponding to the five morphotypes that emerged from the SMS (Figure 3). Shorelines are shown in orange with underlying imagery from Landsat or Google Earth.

The dominant forcings inferred from the delta morphotypes generally align with the estimated relative sediment fluxes driven by each forcing (Figure 5). For example, the river morphotype and wave morphotype deltas lie in the right corners of the Galloway diagram, and the river-wave morphotype deltas span the space in-between these two end member classes with varying degree of relative tidal influence. Note that deltas in the river morphotype typically have relative river sediment flux more than 80%, although there are notable outliers. A similar observation is made for deltas in the wave morphotype. Morphological expression of dominance by a single forcing is therefore limited only to small corners of the Galloway space. Morphologically similar deltas which appear scattered or as misalignments between shoreline-inferred dominant forcing and dominant relative sediment flux in the Galloway diagram yield valuable insight into the relationship between observed shoreline structure and the relative sediment fluxes.

As discussed before, some misalignments arise due to uncertainty in the sediment fluxes estimates. For example, deltas in the tidal morphotype such as the Kolyma and Tigris-Euphrates are assigned relatively low tidal sediment



**Figure 5.** Delta morphotypes on the sediment-fluxed based Galloway diagram. The 54 deltas, colored by their morphotype emergent from the Shoreline Morphometric Space (Figure 3), are positioned in the Galloway diagram based on their estimated relative sediment fluxes (Nienhuis et al., 2020). Misalignments highlight spatiotemporal heterogeneity in the relative sediment fluxes not captured by their contemporary estimates (see text for discussion).

fluxes (Nienhuis et al., 2020), despite displaying clear tidal widening, suggesting under-estimation of the tidal sediment fluxes for these deltas. Similarly, river morphotype deltas such as the Colville, Kuparuk, and Apalachicola, are characterized by abundant mouthbars but have high estimated wave sediment fluxes which are expected to inhibit mouthbar formation (Jerolmack & Swenson, 2007). The Kuparuk and Apalachicola are associated with valley-confined or sheltered shorelines where wave climate data may be particularly uncertain. These misalignments highlight that the shoreline morphometric approach may be more robust than the sediment flux approach for delta classification as it is less sensitive to its defining parameters (e.g., critical angle or range considered for fine scales; see Figure S1 in Supporting Information S1).

Further misalignments of interest are the river-tide morphotype deltas and tide morphotype deltas which are scattered across a range of relative tidal influence. This mixture arises as the river-tide and tide morphotypes consist of deltas with intermediate to high fraction of variance contributed by mouths ( $fM$ ) due to headless and wide channels. However, the river-tide morphotype consists solely of deltas that are convex at the macroscale, for example, the Irrawaddy, Indus, and Mahakam, which is a signature of historical progradation of the delta planform due to fluvial deposition. Also deltas such as the Zambezi and Rufiji are convex with wide headless channels and have abundant tidal mangroves (Anthony et al., 2021; Erfemeijer & Hamerlynck, 2005), suggesting historical significant river and tidal influence, but have otherwise smooth, sandy shorelines and translating spits indicating recent wave influence. This suggests that although these systems at present have large relative tidal sediment fluxes, the estimated relative sediment fluxes do not capture the historical river dominance which constructed them. Thus, as tides widen and preserve former distributary channels (Hoitink et al., 2017), and the timescale for waves to erase the convex depositional system formed by river progradation could be on the order of centuries (Nienhuis et al., 2016), the signature of a river remains on its delta long after it has stopped flowing. Therefore, careful consideration must be given to possible temporal heterogeneities in each of the sediment fluxes when computing their relative values and assessing the relationship between morphotype and relative sediment flux (Bhattacharya & Giosan, 2003). This is especially critical for characterizing morphologic response

to sediment flux changes, for example, decreasing riverine sediment delivery or changes in wave climate, and for projecting delta futures under climate change.

Lastly, we hypothesize that some of the misalignments arise because the morphologic metrics are computed along the length of the entire shoreline, although the sediment fluxes are computed via point estimates and do not reflect information on spatial heterogeneity in the forcings acting on the delta. For example, the Parana delta lies in the wave morphotype although it has a complex distributary network in its southern half and is dominated by riverine sediment flux (Figures 3–5). However, the Parana's depositional environment is unique as the Uruguay river runs parallel to its northern shore (Milana & Kröhling, 2015), which we posit acts as a longshore current that smoothens the shoreface but is not captured by the global sediment flux estimation framework which only includes wind-driven longshore transport. To test this hypothesis, we computed the three multiscale metrics of shoreline structure only on the section of the shoreline between the active distributaries in the southern section, terming it ParanaRO, and found that the ParanaRO indeed lies in the river morphotype (Figure 3), in agreement with its dominant riverine sediment flux (Figure 5).

Note that the multiscale framework presented herein allows us to further interrogate spatially explicit variability in shoreline structure. In particular, some deltas might exhibit lobes corresponding to distinct morphotypes (e.g., abandoned distributary lobes reworked by marine forces following channel avulsion), shedding further light on the alignment between sediment flux and morphology. However, the framework for estimating sediment fluxes (Nienhuis et al., 2020) will likely need to be adjusted to account for highly spatially variable sediment fluxes given multi-lobe or multi-mouth structures or variable wave climate (Syvitski et al., 2022). We note that combining shoreline metrics with metrics of network complexity (Konkol et al., 2022; Tejedor et al., 2015a, 2015b, 2016, 2017) may help to separate deltas further within the SMS and identify subnetworks that need to be treated separately in terms of their morphology and sediment fluxes. Network information may disaggregate the relatively large river-tide morphotypes and the tide morphotypes, with a possible separation of the valley-confined Ob and Yenisei deltas from estuarine systems such as the Kolyma, Ganges Brahmaputra, and Colorado. This further sub-division of deltas may also be able to yield insight into the influence of other controls on delta morphology including grain size (Caldwell & Edmonds, 2014), valley confinement, cold region processes, or sea level history (Nienhuis et al., 2023; Overeem et al., 2022). Interestingly, no systematic signature of near-shore sea-ice, permafrost, or river-ice was detected on shoreline structure (Lauzon et al., 2019; Overeem et al., 2022; Piliouras et al., 2021), except for a lack of wave influenced Arctic systems which may relate to the short wind fetch present due to sea ice (Barnhart et al., 2014) or the presence of a shallow subaqueous ramp dampening wave runup and breakup at the subaerial shoreline (Overeem et al., 2022).

## 5. Conclusion

We have introduced a novel quantitative framework to classify river delta morphology based on a multiscale characterization of delta shoreline structure through geometric and spectral metrics which form a three-dimensional Shoreline Morphometric Space. Unsupervised classification of 54 deltas projected in the SMS reveals self-emergent morphologically similar deltas, that is, delta morphotypes which are further associated with dominant forcings based on the metrics. We then found that dominant forcings inferred from shoreline structure generally align with the dominant forcings quantitatively estimated by their relative sediment fluxes. We posit that misalignments arise due to possible spatiotemporal variability in the dominant forcings not captured in the relative sediment fluxes, providing a basis for more detailed analysis of those deltas. The proposed shoreline morphologic classification framework relies on readily available satellite imagery making it easily applicable for remote, poorly instrumented coastlines and basins as well as on extraterrestrial bodies, for which forcings are not available.

## Data Availability Statement

The values of each metric and relative sediment flux are available in the supplementary material and are available along with code used to reproduce the analysis on Zenodo: <https://doi.org/10.5281/zenodo.7706814>. The Global Surface Water masks used to define the shorelines are available at [https://developers.google.com/earth-engine/datasets/catalog/JRC\\_GSW1\\_1\\_GlobalSurfaceWater?hl=en](https://developers.google.com/earth-engine/datasets/catalog/JRC_GSW1_1_GlobalSurfaceWater?hl=en). ROAM, A fast R-based implementation of the Opening Angle Method, is available at <http://github.com/lvulis/ROAM>.



### Acknowledgments

L.V. was supported by funding from the University of California Lab Fees In Residence Graduate Fellowship Grant L21GF3569, and NASA Earth and Space Science Fellowship Grant 80NSSC18K1409. L.V., A.T., H.M., C.B., and E. F-G., were supported by the United Kingdom Research & Innovation Living Deltas Hub NES0089261. L.V., A.T., H.M., C.B., and E.F-G. were supported by the National Science Foundation through the Collaborative Research program Grant EAR1811909 while D.E., C.B., and J.B. were supported through Collaborative Research program Grant EAR1812019. J.C.R. was supported by the InterFACE project through the DOE. J.H.N. was supported by the Dutch Research Council through Grant NWO-veni.192.123. The authors would like to thank the editor and two anonymous reviewers for their helpful suggestions in improving the manuscript. We also acknowledge helpful discussion on spectral analyses with Phong V. V. Le and Clement Guilloteau, as well as fruitful discussions with the CSDMS community at large.

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