

Concerns about data linking delta land gain to human action

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River deltas are diverse, economically and ecologically important ecosystems that are increasingly vulnerable to environmental change. A recent study reported that global-scale human impact on delta morphology has led to net land area gain (Nienhuis et al.¹). However, we argue that an unconventional definition of deltas, miscellaneous coastal features unduly characterized as deltas and misidentified delta area changes led to spurious statistics in the study by Nienhuis et al.¹ and that a recently published correction² does not address our concerns. We call for the rigorous verification and analysis of large environmental datasets.

Delta identification

Nienhuis et al.¹ defines marine river mouths with fluvial water and sediment discharge thresholds (respectively $>1 \text{ m}^3/\text{s}$ and $> 0.01 \text{ kg/s}$) as deltas. This definition runs counter to standard delta³⁻⁶ geoscience definitions, but nevertheless uses morphological terms that are specific to deltas^{3,4} as distinct from estuaries, such as shoreline *protrusion* and *triangular shape*. Simplified assumptions on river-mouth sediment flux redistribution by river, waves and tides used by Nienhuis et al.¹ neglect the sediment-dispersal role of these agents that render most river mouths devoid of deltas^{3,5,6}. These assumptions also neglect sedimentation pathways interlinking the connected upbuilding and outbuilding components of deltas (subaerial delta-plain, subaqueous delta-front, pro-delta) that differentiate them from simple estuaries.

Nienhuis et al.¹ uses automatically-generated ‘buffer’ areas to identify land change from global spatial data⁷, arguing for the exclusion of areas distant from channel banks and shorelines. We find that this automatic identification of buffers, without verification of their accuracy, leads to thousands of features wrongly identified as “deltas” by Nienhuis et al.¹, including estuaries, built-

37 up areas, rocky coasts (Fig. 1), beach-ridge plains, and multiple distributary mouths of individual
38 deltas. Unfortunately, considering such a collection of *coastal morphologies* as a simple
39 continuum culminating in 10,848 ‘river deltas’ could undermine the gravity of contemporary
40 concerns on delta vulnerability and the hundreds of millions of people deltas host⁸⁻¹⁰. A random
41 check reveals that only ~50% of the inventoried features may be defined as deltas
42 (Supplementary Methods, Extended Data Table 1). Nienhuis et al.¹ checked the veracity of delta
43 existence based on 212 deltas in Madagascar, which is a biased sample. Instead, such a check
44 should be based on a random selection of their large global dataset.

45 **Claimed link between human-impacted river sediment flux and delta land change**

46 Nienhuis et al.¹ conclude that global-scale human impact on river sediment flux has led to delta
47 land gain, which they argue, is illustrated by their Fig. 3a using a *ratio* (Q^d_{river}/Q^p_{river}) of *disturbed*
48 (by dams and land-use changes) and *pristine* (prior to *substantial* human influence) fluvial
49 sediment fluxes, regressed against land change for each delta. Following correspondence with
50 the authors of Nienhuis et al.¹, they affirm not using Q^d_{river}/Q^p_{river} to support their conclusion but
51 the text (p.516) expression $Q^d_{river}-Q^p_{river}$ to impute 16% of delta land gain to human impact.
52 However, Nienhuis et al.¹ does not provide evidence, graphic or otherwise, substantiating this
53 conclusion based on $Q^d_{river}-Q^p_{river}$. We tested $Q^d_{river}-Q^p_{river}$ against land change using their original
54 datasets and methodology and obtained an R^2 of 0.16 (hence their 16%, p.516). However, this
55 value is upheld by the Yellow River delta (Fig. 2), a single delta out of 10,848 ‘river mouths’, and
56 simply considered in their published correction² as an example for which their *methodology may*
57 *produce errors*. Our check shows that this delta’s land area change of -8.3 km²/yr in Nienhuis et
58 al.¹ is flawed (Fig. 2a,d). Our buffer yields +2.15 km²/yr, similar to that published for the same
59 period¹¹, which identifies this delta among the fastest growing on Earth. Whether Yellow River
60 data are corrected for true land change or removed, the R^2 is 0 with $Q^d_{river}-Q^p_{river}$, 0 with
61 Q^d_{river}/Q^p_{river} and 0.09 (p-value: 0.2) for Q^d_{river}/Q^p_{river} using binned data, which is not statistically
62 significant (Fig. 2a, Extended data Fig. 1a). This illustrates the need for caution with particularly
63 influential outliers like this one. Our concern is not only that individual delta change rates are
64 inaccurate, but this inaccuracy permeates into the strong but unsupported conclusion regarding
65 a global-scale link to human-induced sediment flux. This single misidentification led to the
66 incorrect attribution of 16% of land change to human modifications. An $R^2 = 0$ excludes the
67 possibility of a quantifiable link.

68 Errors are recurrent in the automatically-generated buffers of Nienhuis et al.¹, as shown by our
69 analysis of: (1) a random sample of 108 (1%) river mouths and (2) the 100 largest deltas in their
70 dataset (Extended Data Fig. 2b,c). The correlation is very low to nil (R^2 of 0 to 0.13) with our
71 operator-verified buffers. The flawed land change values are not limited to the Yellow River delta.
72 The automatically-generated buffers of Nienhuis et al.¹ are commonly misplaced (hence the

73 plethora of features identified as deltas and deltas with misidentified land change) and/or yield
74 changes neither coastal nor driven by processes claimed by these authors (Extended Data Fig. 2,
75 Extended Data Table 1). The buffer they delimit for the Yangtze, a major world delta, is misplaced
76 by 100 km, yielding land loss instead of strong land gain. Replacing their land change values with
77 those identified from our buffers for the largest 100 deltas in the original (V1) dataset yields bins
78 that are largest for decreased sediment flux ($Q^d_{river}/Q^p_{river} < 1$), revealing the impact of
79 misidentified negative land change in large deltas as shown by the magenta bins in Extended
80 Data Fig. 1a.ii. The land-cover dataset¹² employed by Nienhuis et al.¹ to mask anthropogenic delta
81 transformations is inadequate, leading these authors to grossly overestimate, in populous Asian
82 deltas, land loss (e.g., Mekong, Yangtze, Red River), or gain corresponding to human artefacts
83 (Pearl, Krishna) (Extended Data Fig. 2d-k), such as land conversions into fishponds in the Yellow
84 delta (Fig. 2b), which they recorded as natural *land loss*. They base area-change check on 40
85 deltas in Madagascar and conclude on a standard error of 1%. Madagascar deltas are a biased
86 choice and not a random selection representative of a global-scale dataset. Massive
87 deforestation of this island (44% loss 1953-2014¹³) generates high sediment discharge favourable
88 to delta land gain.

89 There are further inconsistencies in the successive datasets of Nienhuis et al.¹. Following our
90 concerns, these authors generated a new dataset (version 2, V2, March 2021) on the largest 100
91 deltas (Max100), resorting, this time, to more cautious manual, rather than automatic,
92 determination of change. This Max100 dataset shows a 5-fold increase in land area (from 12 to
93 60 km²/yr), but still no statistically significant relationship with the ratio Q^d_{river}/Q^p_{river} or the
94 difference $Q^d_{river}-Q^p_{river}$, nor with the V1 data (Extended Data Fig. 1b,c). This shows that improved
95 buffer definition does not suffice to validate the finding of net delta land *gain* attributed to
96 humans.

97 Nienhuis et al.¹ uses automatic buffers that misrepresent deltas and their changes. Quantifying
98 global delta land changes, especially in large populous deltas, requires clarity in identification and
99 analysis. Deltas are complex features, and examining their global-scale links with human actions
100 faces the challenges of acquiring accurate digital datasets and rigorous modelling. This word of
101 caution is relevant to the analysis of coastal systems and landscape geomorphology faced with
102 increasingly large global datasets, and, hopefully, of interest to a wide audience.

103

104 **Author contributions**

105 All authors developed and contributed to drafts of the text and figures. The concept for this
106 manuscript rebuttal was designed by E.A., F.Z.; F.Z., F.T., M.B. contributed to data analysis and
107 acquisition. All authors (F.Z., E.A., A.V., M.B., F.T.) contributed to the elaboration of the concepts
108 presented here.

109

110 **Competing interests**

111 The authors declare no competing interest.

112

113 **Data availability**

114 We used the original data provided by the corresponding author of Nienhuis et al.¹ by email and
115 data uploaded by the authors of Nienhuis et al.¹ on GitHub at
116 <https://github.com/jhnienhuis/GlobalDeltaChange>.

117

118 **Code availability**

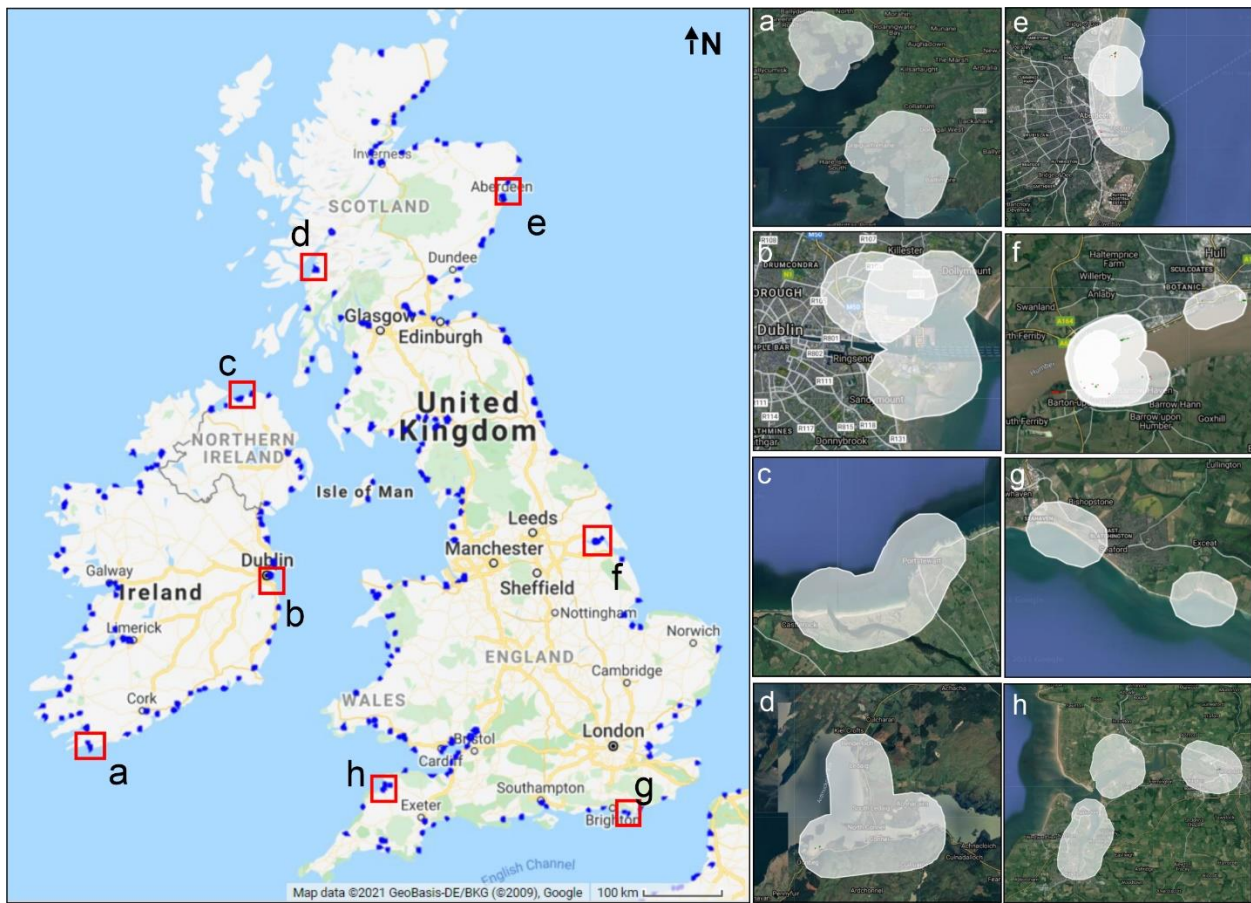
119 All the MATLAB codes used in data analysis and figure production are uploaded in a GitHub
120 repository at https://github.com/FlorinZai/Global_Delta_Check. Interactive Google Earth Engine
121 scripts are also provided at the links in the GitHub repository that can be used to compare our
122 operator-derived buffers with the automatic buffers of Nienhuis et al.¹.

123

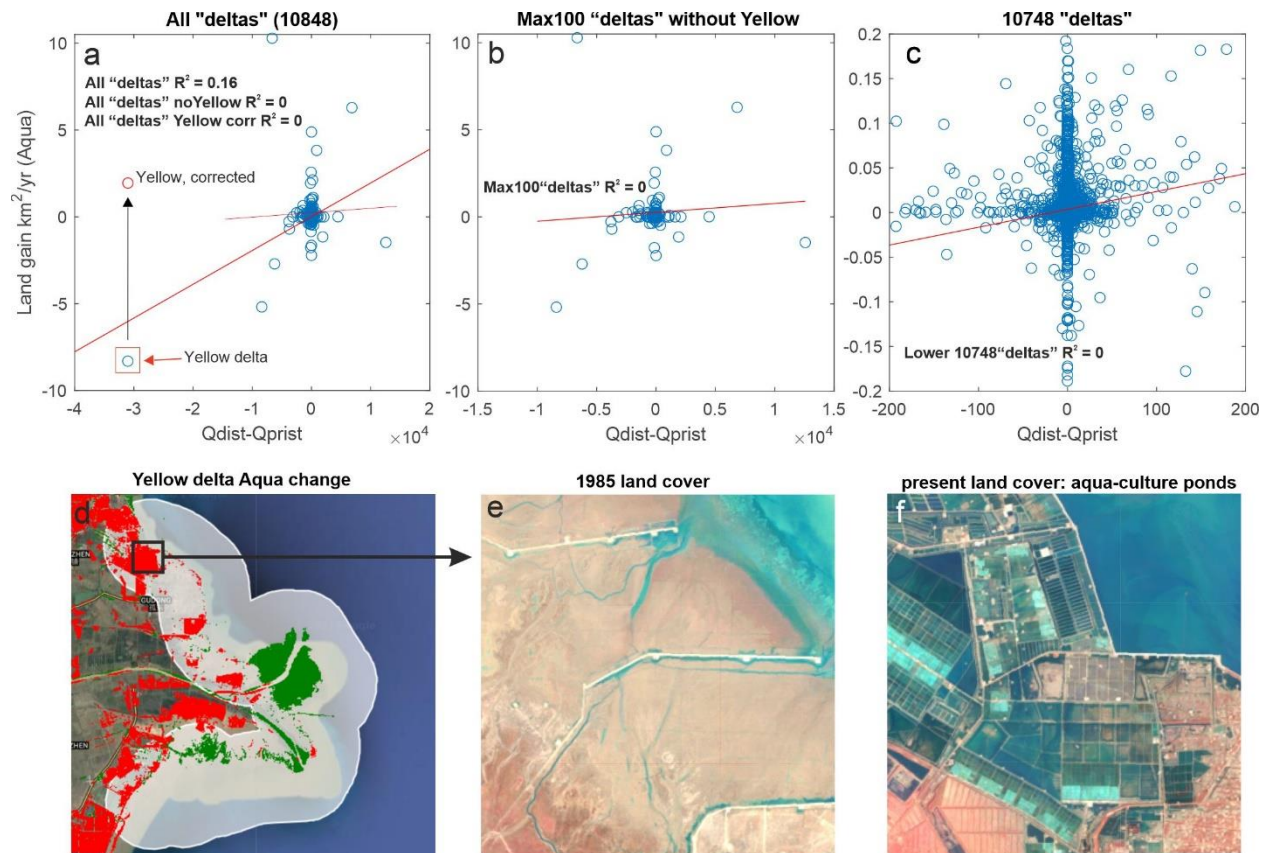
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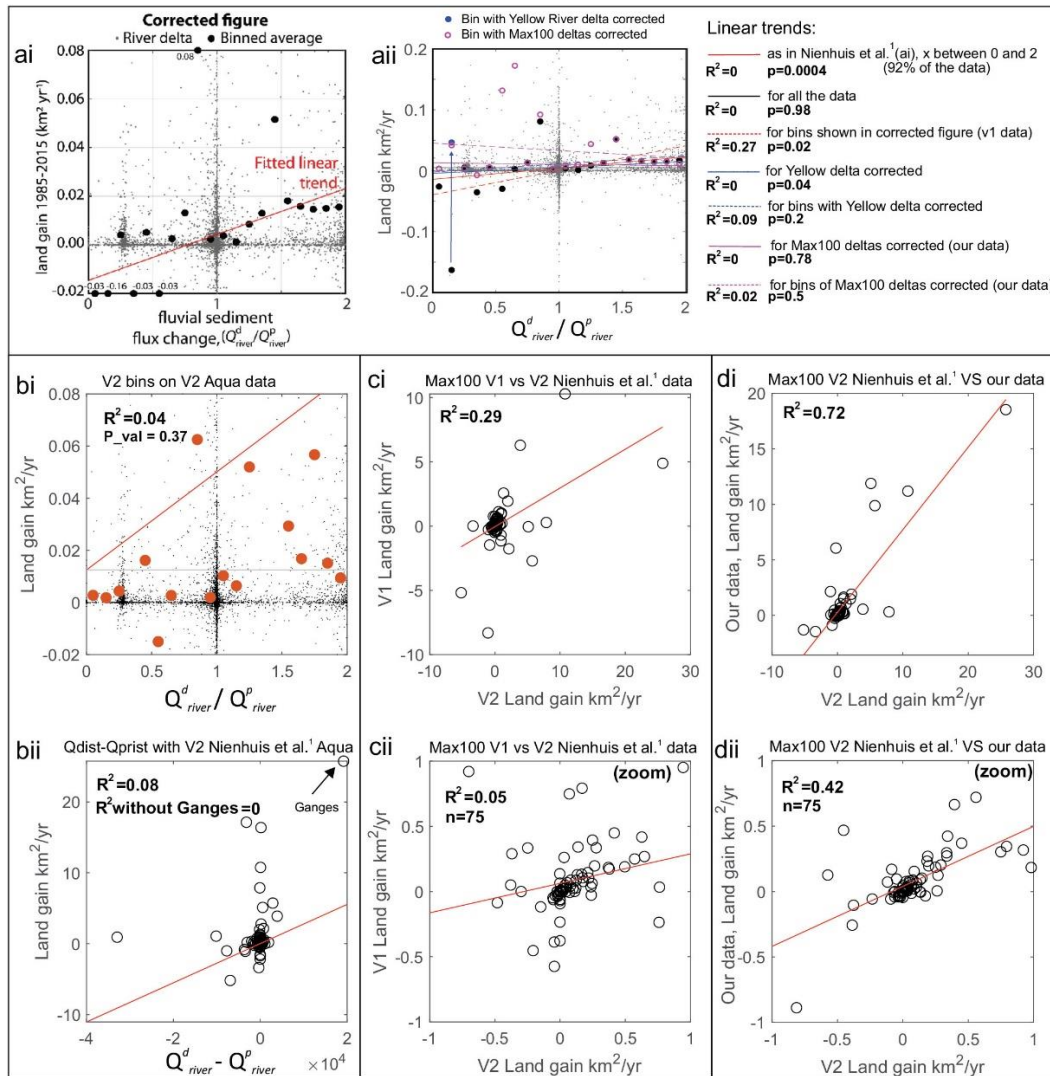


158 **Figure 1.** Spurious categorization by Nienhuis et al.¹ of river mouths as deltas: 270 “deltas” in the British
 159 Isles (blue dots, left), none referenced as a delta in scientific bibliographic databases. Grey zone (a-h)
 160 shows area of each “delta” used to calculate “natural delta” land change by Nienhuis et al.¹, but wrongly
 161 overlain on rocky coasts, beaches, rias, barriers, and cities.
 162



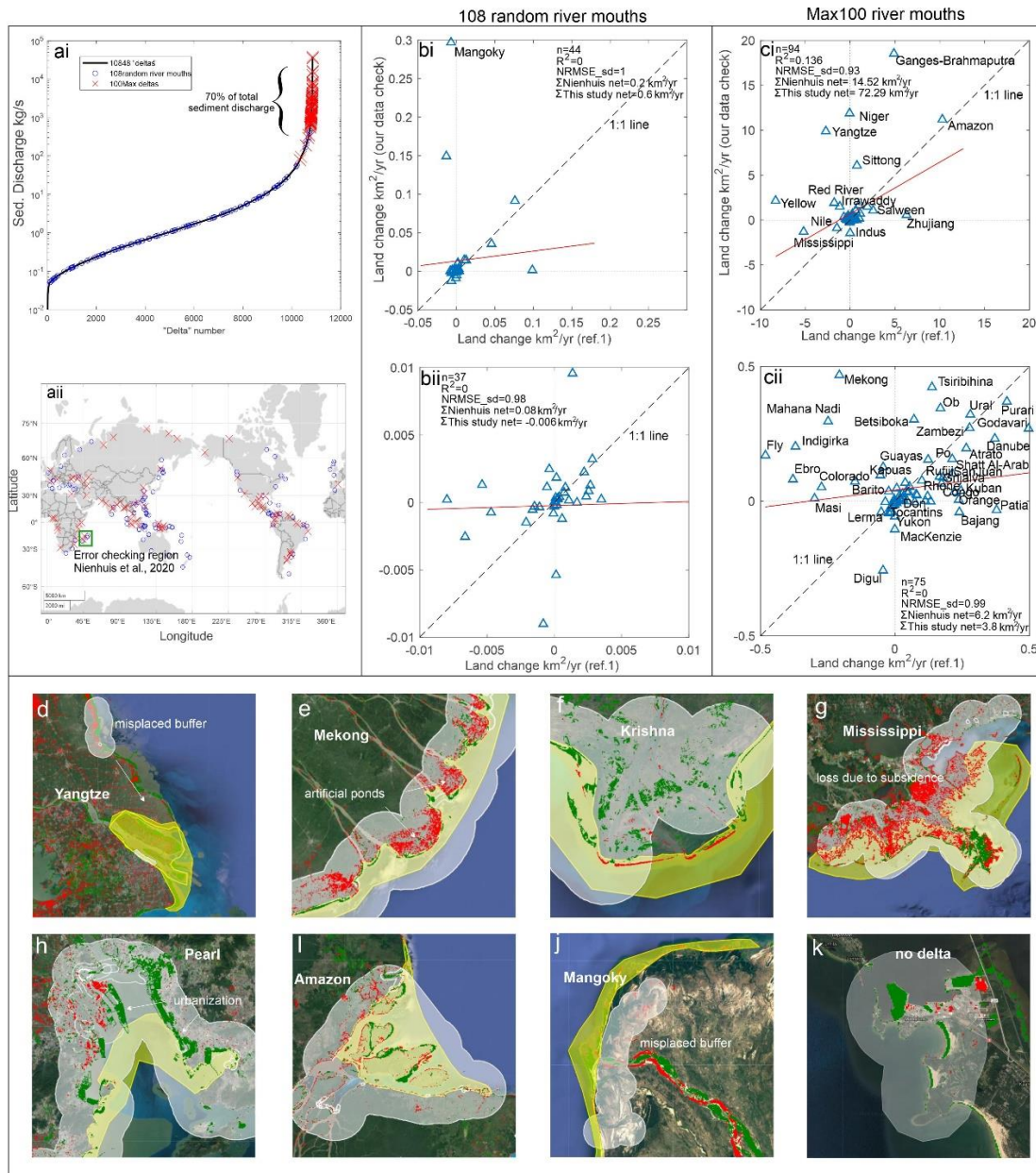
163

164 **Figure 2.** Delta land gain versus human-impacted fluvial sediment flux, $Q_{dist} - Q_{prist}$, shows a weak
 165 correlation, $R^2 = 0.16$, and zero correlation $R^2 = 0$ with the Yellow River delta removed or corrected (a),
 166 and $R^2 = 0$ for both large (b) and small deltas (c). Aquaculture ponds (in red within grey buffer) in Yellow
 167 River delta are misidentified by Nienhuis et al.¹ as natural loss (d). Buffers should capture only natural land
 168 change, not human-transformed lands. Green is land gain.



169

170 **Extended Data Figure 1.** Statistical relationships further invalidating claim of delta land area *gain* due to
 171 Humans. Corrected Fig.3a of Nienhuis et al.¹ (ai) is compared with the same sediment flux change ratio
 172 with different data corrections for the V1 dataset. R^2 of 0 with $p = 0.0004$ is obtained with x between 0
 173 and 2 (92%) of the data. When the whole dataset is considered p becomes non-significant ($p=0.98$). A
 174 correction for the Yellow delta land changes results in R^2 of 0 or very close to 0, and a p-value that is not
 175 statistically significant for bins (aii, blue). Max100 land change (our data) replaced into V1 dataset shows
 176 $R^2 = 0$ and p that is not significant. Continuous lines show regressions for all the 'delta' points; dashed lines
 177 show regressions for the 20 bin averages; axes are extended to show the full variability of bins. Most of
 178 the large deltas fall beyond the chart's y axis. Comparison of the ratio $Q_{river}^d / Q_{river}^p$ with version 2 (V2)
 179 land changes (bi), and of the difference $Q_{river}^d - Q_{river}^p$ with V2 land changes of Nienhuis et al.¹ (bii);
 180 comparison of version 1 (V1, original) and V2 datasets of Nienhuis et al.¹ for Max100 deltas shows low
 181 agreement for all deltas (ci), and no agreement for 75 deltas within $\pm 1 \text{ km}^2$ land change (cii); comparison
 182 of V2 dataset of Nienhuis et al.¹ with Max100 land change obtained through our data check and our
 183 buffers shows good agreement for all deltas (di), and fair agreement for 75 deltas within $\pm 1 \text{ km}^2$ land
 184 change (dii).



185

186 **Extended Data Figure 2.** Distribution of our 108 random samples and the 100 largest deltas (Max100) on
 187 the sediment discharge continuum of Nienhuis et al.¹ (ai), and their global distribution (aai). Note biased
 188 error check of Nienhuis et al.¹ limited to Madagascar Island (aai) (see further comment on the Madagascar
 189 choice in the Supplementary Discussion); comparisons of land changes of Nienhuis et al.¹ with our 108
 190 random samples (44 polygons, bi, bii- zoom on data), and with the Max100 deltas (ci, cii- zoom on data).
 191 Plots bii, cii show subsets of their respective datasets, all with R² = 0. Examples of wrongly identified land
 192 changes by Nienhuis et al.¹ obtained on a selection of large deltas as a result of misplaced buffers (d, j),
 193 human-induced land transformations wrongly reported as natural land changes (e, f, h), and misidentified
 194 deltas (k). Original buffers (light grey) of Nienhuis et al.¹ are compared with our re-drawn data-check
 195 buffers (yellow) based on which only natural, and not human-transformed, delta coastal change should
 196 be calculated.

EE Delta	BasinID	Polygon	Delta	River mouth type	Anthrop.	Anthrop. filtering	Interior change	Small buffer	Buffer overlap	Obs.
0	69	0	1	delta	0	NaN	0	0	0	
1	87861	1	1	delta	0	NaN	1	1	0	River and estuarine dynamics
2	87825	0	0	nondelta	NaN	NaN	NaN	NaN	0	tidal inlet
3	227807	1	1	delta	1	0	0	0	0	Dikes
4	299177	1	1	delta	0	NaN	1	0	0	
5	201340	0	1	anthropogenic river mouth	1	0	0	0	0	Port development, dikes
6	137203	0	0	delta	0	NaN	1	0	0	
7	103071	0	0	river mouth	NaN	NaN	NaN	NaN	0	Dike
8	74573	0	0	interdune channel mouth	NaN	NaN	NaN	NaN	1	No mouth
9	250920	0	0	estuary(ria)	NaN	NaN	NaN	NaN	0	Rocky estuary, river mouth is far
10	135935	0	0	tidal inlets	NaN	NaN	NaN	NaN	0	
11	170497	0	1	delta	0	NaN	1	0	0	
12	157003	0	0	river mouth	NaN	NaN	NaN	NaN	0	No change
13	314898	1	1	delta	0	NaN	0	1	1	Small buffer
14	162388	1	0	delta, anthropogenic river mouth	0	0	0	0	0	Urban development
15	203988	0	0	anthropogenic river mouth	NaN	NaN	NaN	NaN	0	City of Nice
16	2769	1	1	delta	0	NaN	1	0	0	
17	574716	1	1	delta	0	NaN	0	0	0	
18	383601	1	1	delta, anthropogenic river mouth	0	0	0	0	0	Urban development, dikes
19	7818	1	1	delta	0	NaN	1	0	0	
20	276923	0	0	estuary	NaN	NaN	NaN	NaN	0	Dikes, ports, ships, shrimp farms
21	225	0	0	no river mouth	NaN	NaN	NaN	NaN	0	No river mouth
22	151202	0	1	delta	0	NaN	0	0	0	Change from vegetation
23	141533	0	0	river mouth	NaN	NaN	NaN	NaN	0	
24	3018	1	1	delta	0	NaN	1	0	0	
25	673450	0	0	estuary	NaN	NaN	NaN	NaN	0	
26	147976	0	0	estuary	NaN	NaN	NaN	NaN	0	
27	632150	1	1	delta	1	0	0	0	1	Aquaculture, Shrimp farms
28	237383	0	0	estuary(tributary)	NaN	NaN	NaN	NaN	0	No coast
29	119979	0	0	estuary(tributary)	NaN	NaN	NaN	NaN	0	No coast
30	145404	0	0	estuary	NaN	NaN	NaN	NaN	1	
31	27687	0	1	delta	0	NaN	0	0	0	
32	472086	1	1	delta	1	0	1	0	0	
33	382322	1	1	delta	1	0	0	0	0	Urban development
34	124733	0	0	estuary, anthropogenic river mouth	NaN	NaN	NaN	NaN	0	
35	325	0	1	delta	0	0	0	0	0	Dike, change from vegetation
36	91673	0	0	river mouth	NaN	NaN	NaN	NaN	0	Salt ponds, shrimp farms
37	394758	0	0	estuary	NaN	NaN	NaN	NaN	1	estuary
38	51818	0	1	delta	0	NaN	0	0	0	No change
39	689340	0	0	no river mouth	NaN	NaN	NaN	NaN	0	No river mouth
40	295697	1	1	delta	0	NaN	1	0	1	
41	53848	0	0	river mouth	NaN	NaN	NaN	NaN	0	
42	422862	1	1	delta	0	NaN	0	0	1	Vegetation change
43	667218	1	1	delta	0	NaN	0	0	0	
44	738891	0	0	river mouth	NaN	NaN	NaN	NaN	0	No delta, Slope too steep, rocky
45	93	0	0	no river mouth	NaN	NaN	NaN	NaN	0	No river mouth
46	20123	1	1	delta	0	NaN	0	1	0	
47	286863	0	0	river mouth	NaN	NaN	NaN	NaN	0	No delta, Slope too steep, rocky
48	264925	1	1	delta	1	0	0	0	1	Shrimp ponds, dikes
49	249328	1	1	delta	0	NaN	1	1	0	
50	14608	0	0	anthropogenic river mouth	NaN	NaN	NaN	NaN	0	
51	193701	0	0	no river mouth	NaN	NaN	NaN	NaN	0	100% anthropogenic
52	132009	1	1	delta	0	NaN	0	0	0	Few pixels on bar estuary
53	158128	1	1	delta	0	NaN	1	0	0	
54	211649	1	1	delta	0	NaN	0	0	0	
55	199306	0	0	anthropogenic river mouth	NaN	NaN	NaN	NaN	1	Development
56	316765	0	0	no river mouth	NaN	NaN	NaN	NaN	0	
57	262538	1	1	delta	1	0	0	1	0	Anthropogenic ponds
58	718850	1	1	delta	1	0	0	0	0	
59	94741	0	0	no river mouth	NaN	NaN	NaN	NaN	0	
60	412843	1	1	delta	0	NaN	0	1	0	Didn't get spit tip accumulation
61	638519	0	0	river mouth	NaN	NaN	NaN	NaN	1	
62	216515	0	0	river mouth	NaN	NaN	NaN	NaN	0	Fjord?
63	147544	1	1	delta	0	NaN	0	1	0	Small buffer
64	420134	0	0	river mouth	NaN	NaN	NaN	NaN	0	Doesn't look like a delta
65	674874	1	1	delta	0	NaN	0	0	1	River dynamics, coastal works
66	139009	0	1	delta	0	NaN	0	0	1	
67	213083	0	0	delta	NaN	NaN	NaN	NaN	1	Urban development, same delta as #97
68	126813	0	0	no river mouth	NaN	NaN	NaN	NaN	0	No river mouth
69	26952	1	1	delta	0	NaN	1	0	0	
70	145789	1	1	delta	0	NaN	0	0	0	
71	160787	0	0	river mouth	NaN	NaN	NaN	NaN	0	Srtm high coastal elevation
72	25943	0	0	river mouth	NaN	NaN	NaN	NaN	0	
73	145	0	0	no river mouth	NaN	NaN	NaN	NaN	0	No river mouth
74	32721	0	1	delta	0	NaN	0	0	0	Lagoonal delta
75	136526	0	0	river mouth	NaN	NaN	NaN	NaN	1	Barrier diverted river mouth
76	40547	0	0	river mouth	NaN	NaN	NaN	NaN	0	Srtm high coastal elevation
77	71749	1	1	delta	0	NaN	1	1	0	Small & misplaced buffer
78	296274	1	1	delta	0	NaN	1	1	0	
79	48430	0	0	delta	0	NaN	1	0	1	change is River dynamics
80	42223	1	1	delta	0	NaN	0	0	0	Barrier diverted river mouth
81	259998	0	0	river mouth	NaN	NaN	NaN	NaN	0	No delta, Srtm high coastal elevation
82	63338	0	0	tidal inlet	NaN	NaN	NaN	NaN	0	tidal
83	262402	1	1	delta	0	NaN	0	0	1	
84	275141	1	1	delta	0	NaN	0	0	0	
85	34059	0	0	river mouth	NaN	NaN	NaN	NaN	1	River dynamics, Inside bigger delta
86	30270	0	0	delta	0	NaN	0	0	0	
87	395256	0	0	nondelta	NaN	NaN	NaN	NaN	0	seasonal river mouth
88	654755	1	1	delta	0	NaN	0	0	0	
89	634660	0	1	delta	0	NaN	0	0	1	Buffer overlap
90	165985	1	1	delta	0	NaN	1	0	0	Barrier diverted river mouth
91	323297	1	1	delta	0	NaN	0	0	0	jetties at mouth
92	373957	0	0	nondelta	NaN	NaN	NaN	NaN	0	tidal inlet
93	36503	0	0	river mouth	NaN	NaN	NaN	NaN	0	Estuary
94	149	1	1	delta	0	NaN	1	1	0	Small, misplaced buffer
95	331379	1	1	delta	0	NaN	1	0	1	River dynamics
96	39958	0	1	delta	0	NaN	0	0	0	River dynamics, vegetation change
97	213107	1	1	delta	1	0	0	0	1	Urban development
98	598116	0	0	nondelta	NaN	NaN	NaN	NaN	0	No river mouth, tidal creek
99	630689	0	1	delta, anthropogenic river mouth	0	0	0	0	1	Urban development
100	620926	1	1	delta	0	NaN	1	0	0	River dynamics
101	416223	0	0	no river mouth	NaN	NaN	NaN	NaN	0	No river mouth, delta restoration
102	75078	0	0	canals	NaN	NaN	NaN	NaN	1	Aquaculture, shrimp
103	730746	0	0	river mouth	NaN	NaN	NaN	NaN	1	Barrier diverted river mouth
104	583610	0	0	tidal creek	NaN	NaN	NaN	NaN	0	Not delta, not linked to any real river
105	128186	0	0	nondelta	NaN	NaN	NaN	NaN	1	No river mouth, Dikes
106	77603	1	1	delta	1	0	1	0	1	Aquaculture, Inside bigger delta
107	659760	0	0	river mouth	NaN	NaN	NaN	NaN	0	Barrier diverted river mouth
Percentage(%)	41	53			14	0	19	8	25	

197

198 **Extended Data Table 1:** Analysis of 108 randomly selected 'river mouths' from data of Nienhuis et al.¹
199 showing numerous discrepancies. Delta number is EE Delta. Presence is marked 1, absence marked 0; no
200 data due to the absence of a delta is NaN.

201 **Supplementary Methods**

202

203 Nienhuis et al.¹ applies thresholds of water and sediment discharge (respectively $>1 \text{ m}^3/\text{s}$ and $>$
204 0.01 kg/s) to identify river deltas globally, and employs some filtering to exclude small basins (1
205 km^2 for HydroSheds, and either 50 km^2 or 1000 km^2 , based on a drainage divide altitude, for
206 ETOPO1 grid above 60° lat) and deltas draining into fjords. Nienhuis et al.¹ determined land
207 change for each delta from delta extent along the NOAA vectorized shoreline dataset. Nienhuis
208 et al.¹ used Google Earth Engine¹⁴ to retrieve surface-water changes within their selected buffer
209 areas, but noted the potential for sizeable anthropogenic effects and therefore attempts to mask
210 out portions of each delta that are classified as urban/artificial (class 190) by the GlobCover¹²
211 dataset. By selecting only land area change near the NOAA shorelines, Nienhuis et al.¹ claims that
212 land–water conversion within delta interiors is excluded. The delta morphologies identified by
213 Nienhuis et al.¹ are based on quantified river, wave and tidal fluxes, and additional non-quantified
214 criteria reflecting wave and tidal influence deduced from visual observation of features (see their
215 section: “Accuracy of delta morphology prediction”) such as shoreline protrusion and deflection
216 and a triangular shape. These are common plan-view criteria for recognizing deltas as a distinct
217 category of river mouth.

218 We critically examined the dataset and methodology of Nienhuis et al.¹:

219 (i) Firstly, we randomly selected 108 ‘river mouths’, representing 1% of their original dataset of
220 (Extended Data Fig. 2b) to identify delta presence and retrieve land changes. We based this on
221 the afore-mentioned “delta morphology prediction” criteria (including bayhead deltas), and
222 avoiding exclusion where doubt prevailed, but discarding river mouths that had surrounding high
223 elevation based on SRTM data (NASA SRTM Digital Elevation 30 m layer in Earth Engine). From
224 the 108 random river mouth dataset we identify only 57 deltas, the rest being simple estuaries,
225 non-river mouths, etc, but wrongly identified by Nienhuis et al.¹ as deltas (Extended Data Table
226 1). We carefully constructed polygons ($n=44$) only where deltas were present, and based on the
227 criteria of Nienhuis et al.¹, on the deltaic coastal fringes to obtain area changes that are
228 representative of the coastal dynamics of each delta, excluding any interior change. Our
229 operator-based method relies, in line with the essence of this approach, on expert opinion and
230 potentially arbitrary decisions, but no more than their automatic buffer method which also relies
231 on arbitrary delta presence and buffer size definitions. A fundamental difference between the
232 two methods is that an expert operator has the overarching and uncontested advantage of
233 being able to discern whether identified delta land changes are pertinent or not to the processes
234 and forms under scrutiny. We did not draw polygons where deltas had no change or showed
235 overwhelming anthropogenic influence. The remaining deltas are thus not relevant for change
236 analysis and not even one pixel of land change was detected.

237 (ii) Secondly, we checked the 100 river mouths with maximum discharge (Max100) from the V2
238 database (see below) of Nienhuis et al.¹ uploaded on Github
239 (https://github.com/jhnienhuis/GlobalDeltaChange/tree/master/land_area_change). Following
240 correspondence with these authors, they uploaded this new V2 dataset, which retrieves land
241 changes for the 100 river mouths with the largest discharge (Max100), based on manually drawn
242 polygons. These 100 river mouths represent 70% of global sediment discharge (Extended Data
243 Fig. 2c). Out of these, we identified 94 true deltas (94%). Large rivers such as the St. Lawrence,
244 Victoria, Nelson, form estuaries, often bounded by rocky coasts, and the Volga, a true delta,
245 experienced land change caused by water-level variability of the Caspian Sea, and thus cannot be
246 included in this type of study. Area change was obtained based on the same methodology as
247 mentioned above for the 108 random river mouths, first by drawing manually polygons that
248 excluded delta-plain changes representing human occupation, and then retrieving Aqua changes
249 with scripts.

250 Both these datasets (108 random river mouths and Max100 deltas) cover all the spectrum of
251 sediment discharge and are distributed globally (Extended Fig. 2a). Linear regression was applied
252 to both the averaged bins and all the data. We used root-mean-square error (RMSE) and
253 normalized-root-mean-square-error (NRMSE) to quantify errors of the regression model
254 compared to observed values. NRMSE_sd was obtained by dividing the RMSE by the standard
255 deviation of the data.

256 Earth Engine scripts are provided with which users can verify each of the 108 random river
257 mouths (<https://code.earthengine.google.com/1c074d96779371d7bf20922916ca22f9>) and
258 check our Max100 delta land change values and buffers.
259 (<https://code.earthengine.google.com/24d503c385872724f0fc8fe785f2cf42>).

260 **Supplementary Discussion**

261 (i) 108 random samples

262 Regarding the 108 random samples, too many deltas had land change caused by anthropogenic
263 modifications (27%, Extended Data Table 1), indicating unsuccessful filtering by Nienhuis et al.¹,
264 or buffers were misplaced or captured interior delta plain change, and thus leading to an R^2 of 0
265 with their data (Extended Data Fig. 2c,d). In correspondence with the authors, they expressed
266 concerns with our analysis because it is based, they claimed, on a small (<1%), biased subset of
267 their data that does not satisfy basic measures of statistical significance and is limited mostly to
268 small deltas with a combined sediment flux less than 0.5% of the global total. We reiterate that
269 our 1% subset (of which only 0.5% are true deltas) is randomly generated from their global
270 dataset (Extended Data Fig. 2a,b). We also reiterate that Nienhuis et al.¹ did not use random
271 selections from their own dataset to verify affirmations on their global trends.

272

273 (ii) Largest 100 deltas

274 We report our findings relative to the original V1 dataset provided by the corresponding author
275 of Nienhuis et al.¹ by email. Our checks show that there is no agreement between the V1 and V2
276 datasets (Extended Data Fig. 1ci,cii) of Nienhuis et al.¹. In fact, there is more agreement between
277 our manually drawn buffers and the V2 dataset of Max100 deltas (Extended Data Fig. 1di,dii).
278 This actually signifies that the V1 dataset used in the publication of Nienhuis et al.¹ is invalid and
279 confirms our contention that automatic generation of buffers for delta land area change
280 detection is unreliable, as shown by a number of examples from large deltas (Extended Data Fig.
281 2d-k). It is interesting to note that delta lands with limited anthropogenic presence, such as the
282 Amazon, show good agreement (Extended Data Fig. 2i). Hence the approach of Nienhuis et al.¹
283 of using 40 deltas from one single geographic location, Madagascar (Extended Data Fig. 2aii), to
284 check the veracity of delta land areas changes is inappropriate, because anthropogenic
285 modification of deltaic land in this site is virtually nil. Such a check should be based on a random
286 selection of their large global dataset, and not on a statistically untenable criterion of 'coastal
287 continuity' used by Nienhuis et al.¹ to justify the choice of Madagascar. Furthermore, bin
288 averaging (Extended Data Fig. 1a) is inappropriate on a dataset extremely skewed towards large
289 deltas; the first 20 river mouths carry as much sediment as the remaining 10,828. The dataset is
290 noisy, but also has a few extremely large outliers. Any errors in large deltas propagate towards
291 the entire dataset and the bin averages. The authors resort to cutting off the data where
292 $Q_{river}^d/Q_{river}^p > 2$ which showed a $R^2 = 0$ and $p = 0.0004$. Nevertheless, $p = 0.98$, and is therefore
293 not significant when all data are included (Extended Data Fig. 1aii). Clipping off some of the data
294 is not an appropriate approach because normally, a greater increase in sediment flux would result
295 in an even larger land gain signal and a stronger regression.

296
297 Finally, we draw attention to another inconsistency in Nienhuis et al.¹ regarding calculation of
298 delta area. Nienhuis et al.¹ approximated delta area by setting a minimum radius of 2 km for small
299 deltas (1 km shown in the provided codes), adding buffers along the delta shorelines based on
300 delta radius, with an additional 1 km added (shown only in codes, not in paper). To this end, they
301 use in their text the equation of¹⁵: $((1.07 \times Q_{river}^{1.1} \times Q_{w,river}^{0.45} / Dsh(\sim 100) / \pi)^{1/2}$, but then provide a
302 different equation in their Matlab dataset:
303 $((1.07 \times Q_{river}^{0.7} \times Q_{w,river}^{0.45} / \pi) / Dsh(111) \times (\cos(\text{MouthLat})))^{1/2}$. Nienhuis et al.¹ provided no
304 justification for changing the first exponent of Q_{river} from 1.1 to 0.7, for adding a MouthLat
305 cosine bias, and for moving $/Dsh$ after the division with π outside of the area formula.

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