

1 **Evaluating Climate Signals on Global Coastal Shoreline Positions**

2 ***A commentary on “Influence of El Niño on the variability of global***
3 ***shoreline position” by Almar and colleagues (2023).***

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33 **Almar and colleagues¹ are correct in stating that, “*understanding and***
34 ***predicting shoreline evolution is of great importance for coastal***
35 ***management.”* Amongst the different timescales of shoreline change, the**
36 **interannual and decadal timescales are of particular interest to coastal**
37 **scientists as they reflect the integrated system response to the Earth’s**
38 **climate and its natural modes of variability. Therefore, establishing the**
39 **links between shoreline change and climate variability at the global scale**
40 **would be a major achievement. However, we find that the work of Almar et**
41 **al.¹ does not achieve this goal because: (i) the satellite-based method does**
42 **not meet the current standards of practice and produces inaccurate results,**
43 **(ii) the spatial coverage of the shoreline dataset is not adequate for a global**
44 **analysis, (iii) the relevance of the statistical analyses between the shoreline**
45 **data and independent variables is questionable, and (iv) the findings do not**
46 **capture physical patterns of shorelines developed from field-based**
47 **observations.**

48 Here we provide summary information for each of these points:

49 (i) The water line (or “shoreline”) measurement technique of Almar et al.¹ does
50 not follow current standard practices²⁻⁷ including tidal correction, and the
51 resulting data compare poorly with coastal monitoring measurements.
52 Correlation coefficients (r) between the satellite-derived and field-derived
53 shoreline data are only 0.38 to 0.61 (mean = 0.51), suggesting that the
54 technique explains only 14 to 37% (mean = 26%) of the variance of actual
55 shoreline measurements (Fig. S6). That is, almost three-quarters of the

56 variance in actual shoreline positions is not captured by the Almar et al.¹ data.
57 This can be compared to correlations from standard techniques used in other
58 global-scale analyses that capture 67 to 87% (mean = 78%) of the variance of
59 actual shoreline measurements (see Fig. S1 in Vos et al.²)

60
61 This problem is highlighted by the creation of unrealistic seasonality for
62 Narrabeen Beach, Australia (compare thin lines in Fig. S6i) and the failure to
63 capture the largest accretion event of the records, which occurred in 2005 at
64 Torrey Pines, California (compare thick lines in Fig. S6g; Almar et al.¹
65 incorrectly ascribed this event to beach nourishment). If the technique does
66 not demonstratively measure most of the actual shoreline variability,
67 erroneously produces seasonality signals, and fails to capture significant
68 accretion events, then its use to describe global patterns of shoreline
69 variability is likely to lead to inaccurate or misleading results. Almar et al.¹
70 acknowledge that their shoreline results include “hydrodynamic variabilities,”
71 however these variabilities are likely dominated by the confounding effects of
72 tides for many sites⁷. These shortcomings render the technique unsuited for a
73 global analysis.

74
75 (ii) The shoreline dataset includes only one transect every 0.5°, or every 55,000
76 meters on average, to match the resolution of the oceanographic forcing data.
77 This grossly undersamples the world’s shorelines. A single transect at this
78 scale cannot be used to explain regional shoreline patterns because beaches
79 vary greatly in their behavior, orientation, exposure to waves, proximity to

80 sources of fluvial sediment, and human-related modifications. For example,
81 not only can neighboring beaches separated by headlands behave differently
82 due to differences in orientation and exposure, but opposing ends of large
83 embayments can show out-of-phase behaviors due to beach rotation from
84 redistribution of sediment^{8,9}. As such, the spatial undersampling by Almar et
85 al.¹ also increases the chance for spurious results because of the arbitrary
86 location of the transects. In contrast, it is standard practice to map the
87 position of the shoreline at scales of approximately 100 m and employ
88 statistical methods to summarize shoreline behaviors at coarser resolutions as
89 shown by regional and global studies²⁻⁶.

90
91 (iii) Several questions remain regarding the statistical analyses between the
92 shoreline data and the independent variables. The globally averaged
93 correlation (r) of the sea level, wave, and river model was 0.49 (Fig. 1a) and an
94 ENSO-based model resulted in $r = 0.43$ (Fig. 3a). Thus, only ~24% and ~18%
95 of the variance in the shoreline data was explained by these variables.
96 Although these correlations are low but statistically significant, it is important
97 to ask why these correlations exist. Are they a result of actual shoreline
98 changes related to these independent variables? Or are these spurious
99 correlations that result from the inaccurate shoreline data that were not
100 corrected for tidal and other water level effects? To address this, it should be
101 recognized that tides and other water level factors are significantly correlated
102 with ENSO^{2,10,11}, which raises the possibility that the correlations result from
103 residual tidal effects in the shoreline data. In the end, we recognize that ENSO

104 conditions can influence waves, and that waves can strongly influence beach
105 morphodynamics^{2,12}. However, it appears that the statistically significant
106 results of Almar et al.¹ may have been caused by problems in the shoreline
107 data.

108
109 (iv) If a simplified model of global shoreline patterns is to be useful to managers
110 and planners, it should capture the general physical patterns identified by
111 field research and be relevant at the local scale. The results in Almar et al.¹ fail
112 to do this as described in item (i) above. In addition the results suggest that
113 “sea-level dominance” is the overwhelming control on the world’s shorelines
114 (e.g., Central Africa, western Australia, most of the Pacific Rim, and the
115 Mediterranean; Fig. 1b), which contrasts with dozens of scientific studies
116 showing that waves or rivers dominate the morphodynamics at these
117 settings^{2,13,14}. Similarly, river dominance was found for some coastal areas
118 that physically do not receive littoral-grade sediment from their rivers owing
119 to broad, low gradient coastal plains (e.g., much of eastern Australia and part
120 of eastern North America; Fig 1b). These highly unusual results are contrary
121 to the wave dominance that exists broadly for coastal shorelines worldwide¹³.

122
123 Finally, open data are an important element of modern science. Although the authors
124 noted that, “data are made available upon request”, we propose that data publication
125 would have provided a more accessible and lasting option, and it would have been more
126 consistent with Nature’s stated mandates for specific datasets and their endorsement of

127 the Enabling FAIR Data initiative for the Earth, space, and environmental sciences¹⁵.
128 Larger shoreline databases and their processing codes have been published openly, and
129 public-facing viewers of these data serve as good models for getting information to
130 coastal managers and citizens^{2,7}.

131 Combined, the insufficient data and weak correlations suggest that readers should be
132 skeptical of the conclusions from Almar et al.¹ As noted above, many of these
133 conclusions are counter to fundamental understanding of coastal systems, and the
134 general conclusion and headline finding that ENSO is a globally important driver of
135 shoreline change is misleading.

136 We look forward to more rigorous analyses of the trends and causes of coastal change
137 from data that have reasonable uncertainties and are published openly. These kinds of
138 studies are critical to our understanding of coastal habitats and the future of coastal
139 communities during the modern era of human population growth, climate change, and
140 sea-level rise. Coastal managers and citizenry are looking to the scientific community to
141 provide actionable information at both local and regional scales based on rigorously
142 tested and freely available data. Given the importance of this science, future efforts to
143 increase the understanding of coastal systems and carefully reassess the conclusions of
144 Almar et al.¹ will be needed.

145

146 **Competing Interests**

147 None of the authors have competing interests in the publication of this article.

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149 **Author Contributions**

150 JAW led the writing and editing. All others provided background information,
151 intellectual contributions, editing, and/or writing of the manuscript.

152

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