



1 **Title:**

2 Impact of mean sea level rise on storm surges and tide-surge interaction in the German Bight

3

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8

9 **Abstract.** Changes in mean sea level not only directly contribute to extreme water levels (EWLs)
10 but also modulate tidal dynamics, storm surges, and their nonlinear interactions, particularly in
11 shallow shelf seas. This study investigates how sea level rise (SLR) affects storm surges and tide-
12 surge interaction (TSI) in the German Bight, with the aim of assessing the validity of linear
13 approaches commonly used to estimate EWLs in coastal risk studies. A hydrodynamic model
14 was used with a set of controlled experiments for the period 1981–2010, explicitly separating
15 tidal and atmospheric contributions and their interactions, additionally comparing present-day
16 conditions with a +1 m SLR scenario. We demonstrate the importance of tidal phase in
17 modulating the magnitude and timing of storm surges via TSI. The strength of this interaction
18 increases with storm surge height, while its dependence on high tide magnitude and tidal range
19 is comparatively weak. Under SLR, TSI weakens mainly because increased water depth reduces
20 bottom friction, diminishing its dampening effect by about 25 %. Moreover, storm surge heights
21 show a slight decrease, primarily due to phase shifts caused by faster wave propagation in
22 deeper water and reduced wind stress efficiency acting upon a thicker water column. Overall,
23 the nonlinear effects of SLR on storm surges and TSI are small, on the order of a few centimeters
24 for 1 m SLR, relative to the direct contribution of SLR. This supports the use of linear
25 superposition for many practical applications, while highlighting that nonlinear interactions
26 introduce systematic, process-based adjustments relevant for precise EWL estimates.

27



28 **1 Introduction**

29 Extreme water level (EWL) is a combination of many factors, including contributions from storm
30 surges, tides, wave setup, mean sea level (MSL) changes and interactions among those factors.
31 In the North Sea, on short-time scales, storm surges occurring on top of high tides are the
32 primary cause of EWL events. On long-time scales, the observed increase in EWL is mostly
33 associated with the long-term increase in MSL (Haigh et al., 2010; Weisse et al., 2012; Weisse et
34 al., 2014). By 2100, EWL in the North Sea is projected to increase by nearly 1 m under RCP8.5,
35 mainly due to the corresponding rise in MSL (Vousdoukas et al., 2017).

36 According to the IPCC Sixth Assessment Report (AR6), future sea level rise (SLR) in the North Sea
37 will be similar to the global mean projections, with slightly higher rates along the southern
38 North Sea coasts (Fox-Kemper et al., 2021; Garner et al., 2021; Kopp et al., 2023). SLR raises the
39 baseline water level, so even moderate storm surges can trigger extreme impacts and increase
40 the frequency of extremes (Lowe et al., 2001; Hunter, 2010; Taherkhani et al., 2020). Flood risk
41 assessments often account for this by linearly adding SLR to present-day EWL estimates. At the
42 same time, rising water depth alters tides, storm surges and their interactions, especially in
43 shallow regions (Idier et al., 2019). These nonlinear effects remain insufficiently understood and
44 are examined in this study.

45 The shallow coastal areas of the North Sea are known for large tide-surge interactions as well as
46 tide and MSL interactions (e.g. Bruss et al., 2011; Pickering et al., 2012; Idier et al., 2017).
47 Because MSL is rising rapidly and often omitted from EWL projections, it is crucial to determine
48 whether SLR scales linearly with EWL or whether interactions substantially alter this simple
49 relationship. Nonlinear interactions among these components may either amplify or reduce
50 total water levels, depending on the region, tidal phase or storm type (Horsburgh and Wilson,
51 2007; Bruss et al., 2011; Arns et al., 2017; Zhang et al., 2017; Arns et al., 2020). These
52 interactions are essential for understanding water level dynamics and for producing reliable
53 EWL estimates for design and operational forecasting (Zijl et al., 2013; Arns et al., 2017; Arns et
54 al., 2020).

55 Several studies have therefore examined how such nonlinear processes manifest along the
56 North Sea coast. An ensemble study from *Sterl et al.* (2009) showed that no significant change
57 in EWLs on top of SLR along the Dutch coasts. Yet for the shallow areas of the North Frisian
58 Wadden Sea, increases in EWLs beyond SLR were found and attributed mainly to the increase in
59 high tide (Arns et al., 2015).

60 Furthermore, it is useful to estimate how individual components of water level respond to SLR.
61 Many previous studies have investigated tide changes in response to SLR, which have been
62 recognized as spatially incoherent on a global scale (see Idier et al., 2019 and references
63 therein). Even for regional scale, e.g., the North Sea, the results differ substantially (Pickering et
64 al., 2012; Ward et al., 2012; Pelling et al., 2013; Idier et al., 2017). In particular, for the German



65 Bight, studies have reported an increase in the semidiurnal M2 amplitude (Pickering et al.,
66 2012) and high tide levels (Idier et al., 2017), whereas *Ward et al.* (2012) found a decrease in
67 M2 amplitude in response to 2 m SLR. *Pelling et al.* (2013) attributed these differences to the
68 model formulation, specifically whether flooding of low-lying coastal areas is allowed in the
69 model. The newly flooded cells create new shallow areas of high dissipation that counteracts
70 the general decrease of dissipation due to SLR. *Rasquin et al.* (2020) further attributed the
71 substantial discrepancies in the results to the importance of the bathymetry resolution,
72 especially in flat intertidal areas, for modelling the tidal response to SLR.

73 While tidal changes under SLR have received considerable attention, the indirect effect of SLR
74 on storm surges (driven purely by meteorological forcings) remain less well understood. There
75 are studies that use non-tidal residual (e.g. Lowe et al., 2001; Arns et al., 2015) or skew surge
76 (e.g. Sterl et al., 2009) to assess surge changes in response to SLR. Minor changes are found for
77 the shelf region around the United Kingdom and along the Dutch coast (Lowe et al., 2001; Sterl
78 et al., 2009), while a reduction in surge height was reported for the shallow areas of the North
79 Frisian Wadden Sea. There, deeper water reduces the effectiveness of wind stress in dragging
80 water columns and thus leads to the decrease in surge height, despite reduced bottom friction
81 and lower energy dissipation (Arns et al., 2015).

82 Because tides and surges do not evolve independently, their interaction represents a further key
83 process influencing EWLs. Earlier studies suggest that the tide-surge interaction (TSI) is not
84 negligible for reliable EWL predictions (Horsburgh and Wilson, 2007; Olbert et al., 2013; Arns
85 et al., 2020). Its magnitude varies across continental shelves (Arns et al., 2020) and within the
86 North Sea, including the UK coast (Prandle and Wolf, 1978; Horsburgh and Wilson, 2007) and
87 the English Channel (Haigh et al., 2010; Idier et al., 2012). In the North Sea, storm surges
88 typically peak on a rising tide with reduced surge heights due to nonlinear TSI (Prandle and
89 Wolf, 1978). *Horsburgh and Wilson* (2007) attribute this to tidal phase shift caused by faster
90 wave propagation in deeper water and tidally modulated surge generation, with stronger surges
91 at low water because wind stress is more effective in shallow depths. Additionally, local
92 variations in TSI appear across tide gauges, possibly related to different tidal ranges and shape
93 of the coastline. No significant long-term SLR related changes in TSI have been detected in
94 recent decades (Haigh et al., 2010; Feng et al., 2019). This study therefore focuses on TSI in the
95 German Bight and, in particular, its response to future SLR.

96 Given the importance of these nonlinear mechanisms and the remaining uncertainties, our
97 study quantifies the indirect impact of SLR on EWLs in the German Bight, with a particular focus
98 on storm surges and TSI. Rather than providing full projections of EWLs, we concentrate on how
99 SLR modifies these components, in order to assess the validity of commonly applied linear
100 superposition approaches. While TSI describes the nonlinear superposition of tides and
101 meteorologically driven surges under present-day conditions, we additionally investigate how
102 this interaction itself changes when water depth increases due to SLR. To disentangle these

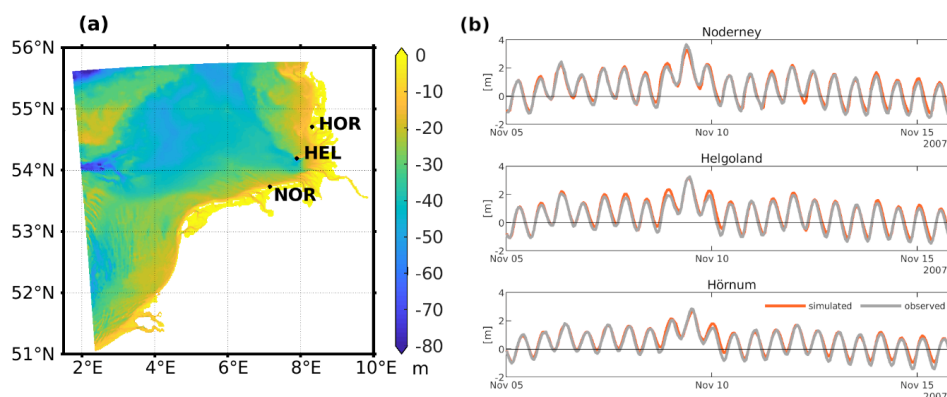


103 processes, we conduct a suite of targeted sensitivity experiments with a hydrodynamic model
104 and complement them with statistical analyses. Section 2 outlines the model setup and
105 experimental design. Section 3 presents TSI and the resulting changes in TSI and storm surges
106 under SLR, followed by a discussion in Section 4. Section 5 summarizes the key findings.

107 2 Model, Definitions and Experiments

108 2.1 Model

109 The experiments used the barotropic hydrodynamic model TRIM-NP (Tidal Residual and
110 Intertidal Mudflat Nested and Parallelized) (Casulli and Stelling 1998; Kapitza, 2008; Pätsch et
111 al., 2017) on a Cartesian Arakawa-C grid allowing flooding and drying of grid points. The bottom
112 friction is parameterized following a depth-dependent Manning-Chezy formulation, and air-sea
113 interaction is represented by bulk parametrizations for the wind components and sea level
114 pressure. The drag coefficient of wind stress is formulated according to *Smith and Banke (1975)*.
115 The model domain covers the North Sea and adjacent parts of the northeast Atlantic with 12.8
116 km spatial resolution, with a nested grid refinement down to 1.6 km in the German Bight. Tidal
117 forcing is applied at the open boundaries using FES-2004 global tidal reanalysis (Lyard et al.,
118 2006), and atmospheric forcing comes from hourly coastDat-2 COSMO-CLM 10-meter wind
119 components and sea level pressure fields (Geyer, 2014). All experiments described below were
120 run for 1981–2010, and model output was stored hourly.



121

122 Figure 1. (a) Model domain and the three selected locations (Helgoland, Norderney and
123 Hörnum) in the German Bight. (b) Modelled (gray) and observed (red) water levels for the event
124 on 9 Nov. 2007 at three stations.

125

126 The model has been previously validated against observations and applied in a wide range of
127 studies (e.g. Weisse et al., 2015; Meyer and Gaslikova, 2024). For the present configuration,
128 model performance was evaluated over the final ten years of the simulation period (2001–



129 2010). The validation was conducted by comparing simulated water levels with tide gauge
 130 observations at three locations along the German coast: Norderney, Helgoland, and Hörnum
 131 (Fig. 1a). These stations are representative for storm conditions in the southern, central, and
 132 northern German Bight, respectively, and are used in the subsequent analysis presented in
 133 this paper.

134 The comparison was performed for hourly water levels, high water levels, and hourly non-
 135 tidal residuals (NTRs). The NTRs were derived by subtracting the tidal signal, estimated using
 136 the U-Tide harmonic analysis Matlab package (Codiga, 2011), from the corresponding
 137 observed or modelled total water levels. Model skill was quantified using the Pearson
 138 correlation coefficient (r), the absolute root mean square error (RMSE), and the RMSE
 139 expressed as a percentage of the maximum observed water level (or NTR) (Table 1), while
 140 Fig. 1b provides a visual comparison of modeled and observed water levels.

141 The model shows good agreement with observations at all three tide gauges for all
 142 considered water level components. For hourly water levels, Pearson correlation coefficients
 143 are high, ranging from 0.97 at Hörnum to 0.98 at Helgoland, with RMSE values between
 144 0.2 m and 0.23 m, corresponding to 6 % of the maximum observed water level. High water
 145 levels are also well reproduced, with correlation coefficients close to 0.9 at all stations and
 146 comparatively low RMSE values (0.16 m), amounting to 4–5 % of the highest observed
 147 levels. Performance for the hourly NTRs is also robust, with correlations between 0.86 and
 148 0.89 and RMSE values of about 0.16 m (5–6 % of the maximum observed NTR). Overall, the
 149 validation results indicate a consistently high model skill across the German Bight, with only
 150 moderate spatial variations between the southern (Norderney), central (Helgoland), and
 151 northern (Hörnum) stations. The discrepancies can be partly attributed to uncertainties in
 152 the atmospheric forcing, as discussed in *Meyer and Gaslikova (2024)*, and in part to the
 153 relatively coarse horizontal resolution of the model (1.6 km), limiting the performance
 154 especially in the wide shallow areas and intertidal flats. The present model configuration
 155 therefore represents a deliberate trade-off between accuracy and computational efficiency,
 156 which is necessary to enable the execution of multiple multi-decadal simulations.

157

158 Table 1: Comparison with tide gauge observations for 2001–2010.

	water level			high water			non-tidal residuals		
	r	RMSE (m)	RMSE (%)	r	RMSE (m)	RMSE (%)	r	RMSE (m)	RMSE (%)
Helgoland	0.98	0.21	6.4	0.90	0.16	4.9	0.88	0.16	6.2
Norderney	0.97	0.23	6.1	0.90	0.16	4.3	0.86	0.16	5.3
Hörnum	0.97	0.20	6.1	0.90	0.15	4.6	0.89	0.16	5.5

159



160 2.2 Key terms and definitions

161 To ensure clarity in interpreting the results, key terms used throughout the text are defined
162 below.

163 **Water level (WL)** denotes the observed or total water level, accounting for all contributing
164 components and their interactions.

165 **Tide (TIDE)** refers to the regular rise and fall of water level caused primarily by the gravitational
166 pull of the Moon and the Sun.

167 **Storm surge (SURGE)** is the direct response of water level to strong winds and low atmospheric
168 pressure associated with storms.

169 **Tide-surge interaction (TSI)** describes the nonlinear interplay between tides and storm surges. It
170 can raise or lower the total water level relative to their linear superposition, especially in
171 shallow coastal areas.

172 **Non-tidal residual (NTR)** represents the combined effects of all non-tidal processes. It is often
173 approximated as the difference between the observed water level and the predicted tide,
174 representing the combined effect of storm surge and TSI (Idier et al., 2019).

175 **Effects of SLR:** SLR modifies tidal and surge dynamics through depth-dependent processes.
176 Three SLR-related effects are considered:

- 177 • **Impact of SLR on storm surge (ISS)**
- 178 • **Impact of SLR on tides (IST)**
- 179 • **Impact of SLR on tide-surge interaction (ISTSI)**

180 They quantify how SLR alters surge height, tide and their nonlinear interactions, respectively.

181

182 2.3 Experimental set-up

183 To separate the contributions of tide, storm surge and their interactions, the model was run in
184 three configurations: **FULL** runs with tidal and atmospheric forcing, **TIDE** runs with tidal forcing
185 only, and **SURGE** runs with atmospheric forcing only. To investigate the potential impact of
186 future SLR, each configuration was simulated under two sea-level conditions: **0 m SLR (..._0)** and
187 **+1 m SLR (..._1)**. SLR of 1 m was chosen to represent a plausible high-end scenario consistent
188 with IPCC AR6 global projections of 0.77 (0.63–1.01) m under SSP5-8.5 by 2100 and regional
189 estimates of 0.85 (0.61–1.61) m for Cuxhaven (Fox-Kemper et al., 2021; Garner et al., 2021;
190 Kopp et al., 2023), as well as other high-end assessments exceeding 1 m (Katsman et al. 2011).
191 In total, six experiments were conducted: **FULL_0**, **FULL_1**, **TIDE_0**, **TIDE_1**, **SURGE_0**, and
192 **SURGE_1** (Table 2). For example, **SURGE_1** represents the surge-only run under +1 m SLR and
193 provides the wind-driven surge together with its interaction with the elevated MSL.



194

195 **Table 2:** Sensitivity Experiments.

EXPERIMENT ID	FORCING INCLUDED	SLR CONDITION	PURPOSE / OUTPUT
FULL_0	Tide + Atmosphere	0 m	Baseline total water level (WL)
FULL_1	Tide + Atmosphere	+1 m	WL under SLR (WL_SLR)
TIDE_0	Tide only	0 m	Baseline tidal signal (TIDE)
TIDE_1	Tide only	+1 m	Tide under SLR (TIDE_SLR)
SURGE_0	Atmosphere only	0 m	Baseline storm surge (SURGE)
SURGE_1	Atmosphere only	+1 m	Storm surge under SLR (SURGE_SLR)

196

197 **2.4 Computation of residual effects and interaction terms**

198 The following diagnostic quantities were derived from the model simulations defined above.

199 **Tide-surge interaction (TSI)**

200
$$TSI = WL - TIDE - SURGE \tag{1}$$

201 Positive *TSI* values indicate constructive interaction (amplification), while negative values
202 indicate destructive interaction (reduction).

203 **Non-tidal residual (NTR)**

204
$$NTR = WL - TIDE = SURGE + TSI \tag{2}$$

205 It should be noted that NTR is not equivalent to storm surge in the presence of TSI, as it also
206 includes the nonlinear interaction term.

207 **Impact of SLR on storm surge (ISS)**

208
$$ISS = SURGE_SLR - SURGE - SLR \tag{3}$$

209 This definition isolates the dynamical response of storm surges to SLR, independent of the
210 direct increase in MSL. Negative values indicate that surge height decreases under SLR, such
211 that the water level increase is less than the SLR itself. Positive values indicate that surge height
212 increases, leading to water level changes exceeding SLR alone.

213 **Impact of SLR on tides (IST)**

214
$$IST = TIDE_SLR - TIDE - SLR \tag{4}$$

215 Analogously, this definition quantifies the dynamical response of the tidal signal to SLR,
216 excluding the direct effect of the imposed sea-level change.



217 **Impact of SLR on tide-surge interaction (ISTSI)**

218
$$ISTSI = WL_{SLR} - WL - SLR - IST - ISS \quad (5)$$

219 Positive values indicate higher TSI values under SLR, while negative values indicate a decrease in
220 TSI values.

221

222 **3 Results**

223 Storm surge events were identified using the SURGE_0 experiment, which serves as the baseline
224 for the analysis. This choice ensures that event selection is independent of tidal phase and
225 allows a consistent analysis of phase-dependent interaction effects. The 98th percentile of surge
226 peaks was used as a threshold, and events were declustered to ensure at least 24 hours of
227 separation. The number of detected events varies across locations, with 240, 224, and 213
228 events detected at three different locations along the German coast: Norderney, Helgoland, and
229 Hörnum, respectively (Fig. 1a). For consistency, the 200 largest events were analyzed at each
230 site, unless otherwise specified. As the results were similar across locations, only the analysis for
231 Helgoland is presented in the paper.

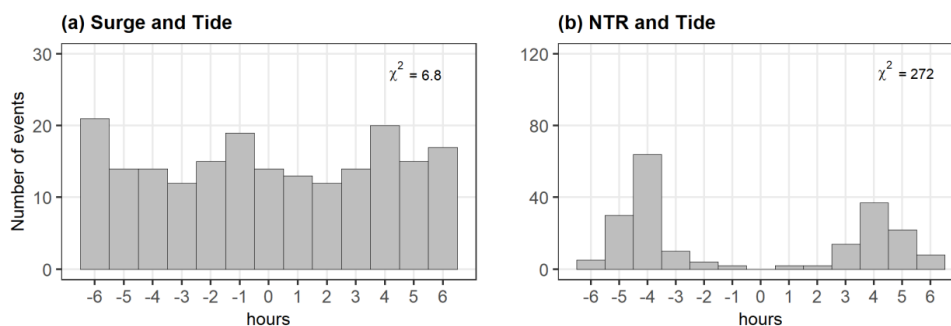
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233 **3.1 Tide-surge interaction (TSI)**

234 To assess whether TSI occurs in the region, we first compared the distributions of surge peaks
235 with and without interaction. Figure 2a shows the timing of surge peaks relative to high tides,
236 using surge peaks from the surge-only run (SURGE_0) and the nearest high tides from the tide-
237 only run (TIDE_0). In this configuration, the two runs are independent, so surge peaks should be
238 uniformly distributed over the tide cycle. This is indeed observed: the peaks are almost evenly
239 spread over all hourly bins, and the chi-square test statistic remains well below the critical
240 value, confirming no deviation from a uniform distribution. The test statistic χ^2 is computed as
241 (e.g. Haigh et al., 2010):

242
$$\chi^2 = \sum_{i=1}^n [(N_i - e)^2 / e], \quad (6)$$

243 where N_i represents the number of observed events per time bin i , e is the number of expected
244 events per bin under the null hypothesis of a uniform distribution, and n represents the number
245 of bins (here 13 hourly bands were used). Deviation from a uniform distribution is considered
246 significant at the 95 % level if the test statistic exceeds the critical chi-square value ($\chi_{12,0.95}^2 =$
247 21), where 12 is number of degrees of freedom.

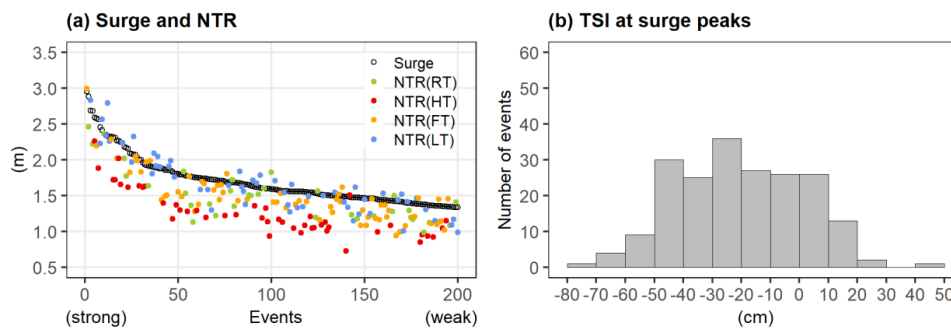


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249 Figure 2. Frequency distribution of (a) surge peaks and (b) non-tidal residual (NTR) peaks
 250 relative to the timing of high tide. The corresponding chi-square statistics are shown in each
 251 panel.

252

253 When NTR peaks were used instead of surge peaks, the distribution is markedly different (Fig.
 254 2b). The NTR peaks occur most frequently on the rising tide, about 4–5 hours before high tides.
 255 This is in agreement with previous studies in this region (Horsburgh and Wilson, 2007; Sterl et
 256 al., 2009; Idier et al., 2012) and in other shallow regions of the world (Zhang et al., 2017; Koh et
 257 al., 2024), which likewise found residual peaks tending to avoid high tides. *Horsburgh and*
 258 *Wilson* (2007) attributed this to phase shift of the tidal signal caused by surge-induced depth
 259 changes and to the modulation of surge production by water depth. The contrast between
 260 Figs. 2a and 2b demonstrates that NTR is not simply the water level response to atmospheric
 261 forcing and that TSI plays a crucial role in shifting the timing of surge peaks. This further
 262 confirms that NTR cannot be interpreted as a direct measure of storm surge, as it also contains
 263 the contribution from TSI.



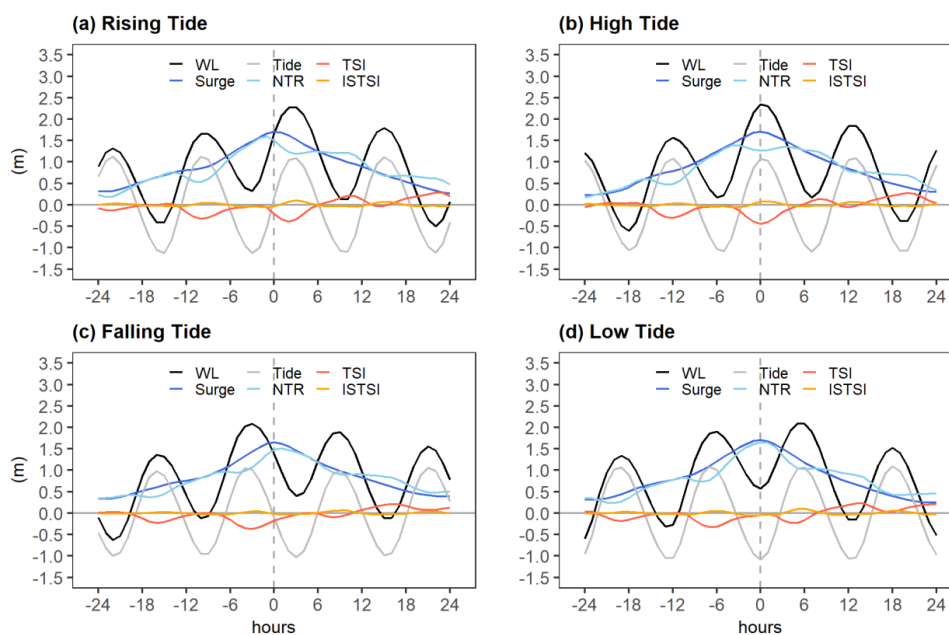
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265 Figure 3. (a) Comparison of surge (black open cycles) and NTR (colored points) at surge peaks,
 266 ordered from highest to lowest surge. The NTR values are grouped by tidal phase: rising tide (RT,
 267 rising tide (RT, green), high tide (HT, red), falling tide (FT, orange), and low tide (LT, blue). (b) Frequency
 268 distribution of TSI at surge peaks.



269

270 Figure 3a shows that surge height and NTR height are only weakly linked. While NTR values
271 generally decrease with decreasing surge height, the NTR values across tidal phases show
272 substantial scatter. NTR at high tide (red) tends to lie near the lower end of the range, while NTR
273 at low (blue), rising (green), and falling tide (orange) more often show higher values. The
274 comparison of surge height and NTR at surge peaks further shows that NTR is generally lower
275 than the corresponding surge height (Fig. 3a), indicating a reduction of water level by TSI at
276 surge peaks. On average, NTR is about 12 % lower than the corresponding surge maxima. This is
277 consistent with the distribution of TSI at surge peaks (Fig. 3b), which is predominantly negative,
278 with values mostly between -54 cm and 14 cm (5th-95th percentiles) and a mean of about
279 -21 cm.



280

281 Figure 4. Event-averaged curves for water level (black), tide (gray), surge (blue), NTR (light blue),
282 tide-surge interaction (TSI, red), and the SLR impact on TSI (ISTSI, orange), shown around the
283 surge peak (gray dashed line) for surge peaks occurring during (a) rising tide, (b) high tide, (c)
284 falling tide and (d) low tide.

285

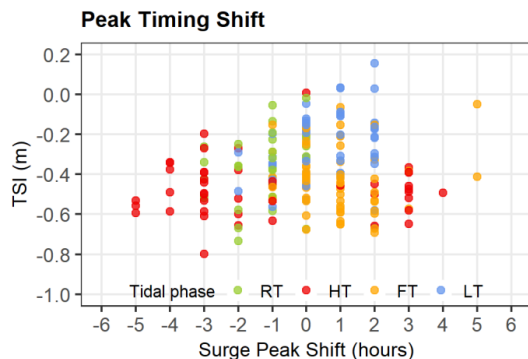
286 To examine the phase dependence more systematically, events were grouped by the tidal phase
287 at which the surge peak occurred, and event-averaged curves were constructed (Fig. 4). This is
288 possible because the events are identified with the surge-only run and are therefore not
289 influenced by tidal phase. High and low tides are defined as the tidal maximum and minimum,



290 including a ± 1 hour interval around each. At Helgoland, for example, 33 events occurred at
 291 rising tide, 48 at high tide, 62 at falling tide, and 57 at low tide. Across all groups, tide and TSI
 292 are generally negatively correlated (gray and red curves), consistent with the behavior described
 293 in *Song et al. (2020)* for Bohai Bay in northeastern China. The strongest negative TSI occurs
 294 mostly near high tide or on rising tide rather than at surge peak. This is in line with previous
 295 studies, showing that the maximum TSI effect does not necessarily coincide with the maximum
 296 surge (Rego and Li, 2010; Zhang et al., 2017; Huang et al., 2025).

297 The negative TSI indicates a dampening effect on surge heights, which is also evident from the
 298 comparison between surge and NTR (blue and light blue curves). The NTR values are generally
 299 lower than the surge peaks, with the magnitude of the reduction depending on tidal phase. TSI
 300 is typically negative during rising surges, reflecting a consistent reduction of surge height, but it
 301 can become positive during falling surges, when the interaction can briefly enhance water
 302 levels.

303 The strongest reduction occurs when surge peak coincides with high tide, causing NTR to reach
 304 its maximum before or after surge peaks (Fig. 4b). This phase also shows the largest timing
 305 shifts, with two clusters around ± 3 hours (red points in Fig. 5). When surge peaks occur at low
 306 tide (Fig. 4d), the interaction is weak, and NTR peaks lie within about one hour of surge peaks
 307 (blue points in Fig. 5). The rising- and falling-tide cases show intermediate reductions (Figs. 4a
 308 and 4c), with typical timing shifts of about -1 hour during rising tide and $+1$ hour during falling
 309 tide (green and orange points in Fig. 5). Similar phase-dependent nonlinear effects were
 310 reported for Hurricane Rita landfall timing experiments, where low-tide landfalls enhanced and
 311 high-tide landfalls reduced total water levels (Rego and Li, 2010).



312

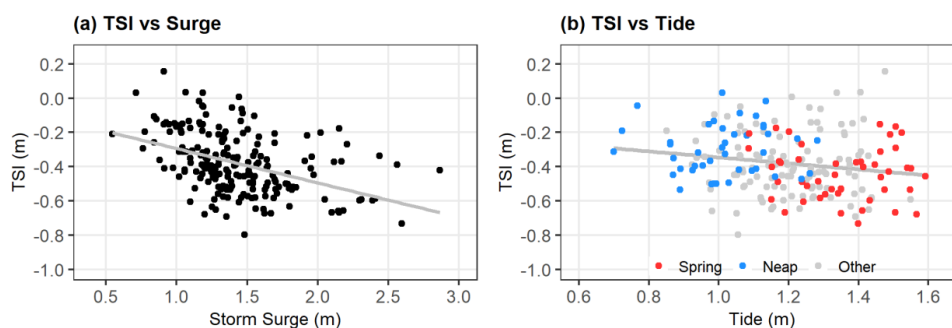
313 Figure 5. TSI at high tide versus shift in peak timing between NTR and surge ($PT_{NTR} - PT_{Surge}$),
 314 conditioned upon tidal phase: rising tide (RT, green), high tide (HT, red), falling tide (FT, orange)
 315 and low tide (LT, blue).

316



317 The curves in Fig. 4 represent averages across subsets of events. However, the events in each
318 subset show considerable variability in terms of surge and NTR relations. For events occurring
319 near low tide (Fig. 4d), the averaged surge and NTR are close to each other at surge peak, yet
320 more than 40 % of the low-tide cases exhibit positive TSI, meaning NTR exceeds surge (blue
321 points above black cycles in Fig. 3a). From the perspective of the positive TSI events, over half of
322 the events occur when surge peak is near low tide, with the remainder split roughly evenly
323 between rising and falling tides. In contrast, for cases near high tide (red points in Fig. 3a), NTR
324 is consistently lower than surge, confirming that TSI reaches its strongest negative values
325 around high tide and exerts a clear dampening effect on surge maxima. Overall, the results
326 demonstrate that the magnitude and timing of TSI effects are strongly dependent on tidal
327 phase.

328 We next examine how surge height and tidal range influence TSI. When TSI values are
329 considered at all high tides, regardless of the presence of surge, the distribution is roughly
330 symmetric around zero (not shown). The negatively biased distribution of TSI seen in Fig. 3b
331 appears only under surge conditions. As also shown above, TSI is strongly dependent on tidal
332 phase. To condense the results in the following analysis, the same 200 events (Section 3) are
333 therefore evaluated at the moment of high tide rather than at surge peaks. This approach
334 captures the strongest negative TSI and provides a more relevant basis for EWL estimates,
335 whose peaks typically occur near high tide. As shown in Fig. 6a, TSI becomes more negative as
336 surge height increases, indicating that the dampening effect of TSI tends to strengthen during
337 stronger surges. This is consistent with the expectation that stronger surges generate stronger
338 currents and thus stronger nonlinear interactions, as also noted by *Zhang et al.* (2017).



339

340 Figure 6. (a) Scatter plot of TSI versus surge height at high tide with fitted regression line. (b)
341 Same as (a), but for the high tide magnitude. Points occurring during spring and neap tides are
342 highlighted in red and blue, respectively.

343

344 The magnitude of high tide, defined here as the instantaneous water level at high water in the
345 tide-only run (TIDE_0), and tidal range in general also influence TSI, though less clearly.

346 Compared to the trend for surge height, the dependency of TSI on high tide magnitude is



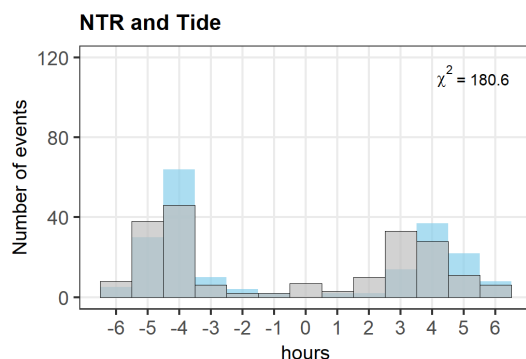
347 weaker and more scattered (Fig. 6b). High tide magnitude should be distinguished from tidal
348 range, which describes the difference between high and low water and is more directly related
349 to spring-neap variability. When the events are separated into spring and neap tides, TSI is
350 generally stronger during spring tides and weaker during neap tides (mean values of -44 cm and
351 -30 cm, respectively), but with a substantial overlap between the groups. This overlap reflects
352 that high tide magnitude and tidal range are not uniquely related in the data, so that higher
353 neap tides may exceed lower spring tides in individual cases. Similarly, *Tien et al.* (2025) also
354 reported higher surge (equivalent to NTR in the present study) during neap tides and lower
355 surge during spring tides in idealized typhoon experiments.

356

357 3.2 Impact of SLR on tide-surge interaction (ISTS)

358 We next investigate how TSI changes under SLR. Previous studies have assessed the temporal
359 changes of TSI by comparing different decades using tide gauge observations. While no clear
360 long-term change in TSI distribution was found (Haigh et al., 2010; Feng et al., 2019), minor
361 changes have been noted (Feng et al., 2019), such as a shift in peak timing towards high tide at
362 one gauge and slightly higher NTR at another. These subtle changes may indicate a minor
363 weakening effect of SLR on TSI, though they are not consistent across locations and could be
364 affected by decadal MSL and surge variabilities.

365 SLR effects on TSI could be quantified with our simulations. Computing NTR from the
366 experiments under SLR (Table 2, Eq. (2)), Fig. 7 shows nearly unchanged distribution patterns
367 compared to Fig. 2b, although the first peak becomes flatter and the second shifts one hour
368 towards high tide. Although the chi-square test still confirms statistically significant deviation
369 from the uniform distribution, the chi-square value decreases slightly, indicating that TSI persists
370 but it weakens under 1 m SLR.

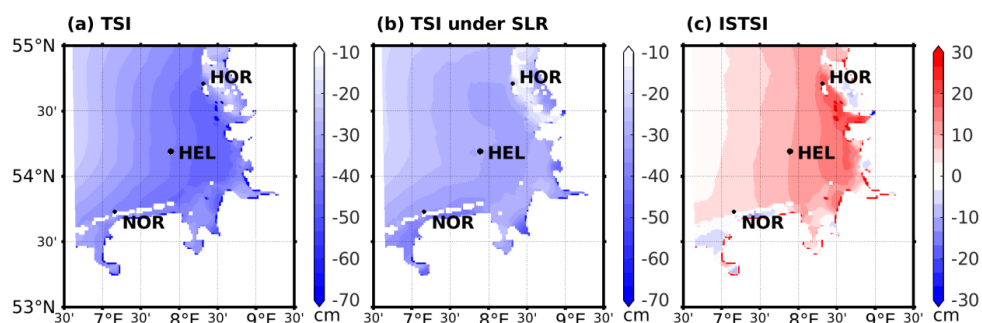


371

372 Figure 7. Frequency distribution of NTR peaks relative to the timing of high tide under 1 m SLR
373 (gray bar). The light blue bar in the background is the distribution under no-SLR condition from
374 Fig. 2b.



375



376

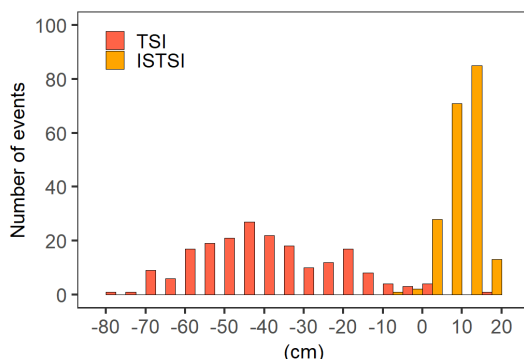
377 Figure 8. Maps of (a) TSI with no SLR, (b) TSI under 1 m SLR and (c) the SLR impact on TSI (ISTSI)
 378 at the nearest high tides of the surge events detected at three locations within a 1 h time
 379 window.

380

381 As can be inferred from Fig. 4 (orange curve), ISTSI is generally inversely related to the original
 382 TSI. When TSI reaches its minimum near high tide, SLR induces a positive TSI change, indicating
 383 a weakened nonlinear effect. This is a basin-wide effect, as is evident from Fig. 8. There the TSI
 384 under present-day and +1 m SLR conditions is estimated based on surge events co-occurring
 385 throughout the German Bight and detected at three locations within a 1 h time window (87
 386 events). The event-averaged TSI becomes less negative under SLR compared to the control case,
 387 demonstrating a systematic weakening of TSI. This behaviour can be largely explained by the
 388 reduction of bottom friction associated with increased water depth. Reduced friction leads to
 389 lower energy dissipation, which tends to increase local water levels, while simultaneously
 390 weakening the nonlinear momentum exchange between tidal currents and surge flows. As a
 391 result, the coupling between tides and surges is reduced, and the system response becomes
 392 more linear (Idier et al., 2019).

393 The magnitude of this weakening is, however, small. Around high tide, TSI ranges from –65 cm
 394 to –9 cm (5th–95th percentiles) with only five positive cases and a mean value of about –38 cm
 395 (Fig. 9), consistent with values reported previously (Rego and Li, 2010; Song et al., 2020; Huang
 396 et al., 2025). More positive cases exist for TSI at surge peaks (Fig. 3a), because the effect is
 397 weaker at surge peaks. The ISTSI around high tide mostly lies between 2 cm and 15 cm, with
 398 only three negative cases and a mean of 10 cm. Thus, it reduces the original TSI by
 399 approximately 25 % on average.

400



401

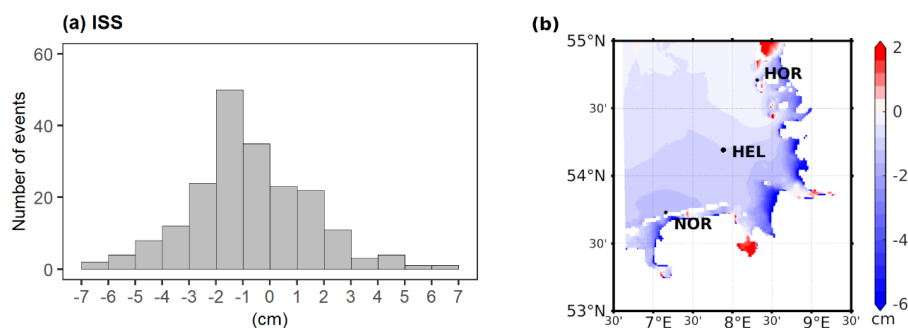
402 Figure 9. Frequency distribution of TSI with no SLR (red) and ISTSI (orange) at high tide.

403

404 3.3 Impact of SLR on storm surges (ISS)

405 The impact of SLR on storm surges is estimated from the two surge-only runs. The same wind
406 forcing was used for both simulations, so their differences stem only from the change in MSL. In
407 other words, the surge response reflects only SLR, independent of tides or atmospheric
408 variations. Surge changes are determined as differences between corresponding peaks of the
409 control run (SURGE_0) and the run with 1 m SLR (SURGE_1) as shown by Eq. (3). Surge changes
410 due to SLR show an asymmetric distribution dominated by negative values (70 %), indicating on
411 average a slight reduction in surge height in response to SLR (Fig. 10a). Figure 10b shows the
412 spatial distribution of ISS for the coincident surge events detected at three locations within a 1 h
413 time window, highlighting a basin-scale reduction in relative surge height under higher MSL. The
414 reduction reflects a combination of two opposing depth-related effects as SLR increases total
415 water depth: while this leads to reduced bottom friction and thus less dissipation, which tend to
416 promote higher surge development, it also makes surface wind stress less effective by acting on
417 a thicker water column, thereby dampening surge generation (e.g. Arns et al., 2017). The
418 strongest negative response appears about 5 hours after the surge peak (Fig. 11), indicating a
419 depth-induced phase shift caused by faster wave propagation in deeper water. As a result, surge
420 height evaluated at the surge peak timestep is also slightly lower, although the difference is
421 small and not easily visible in the surge curves in Fig. 11.

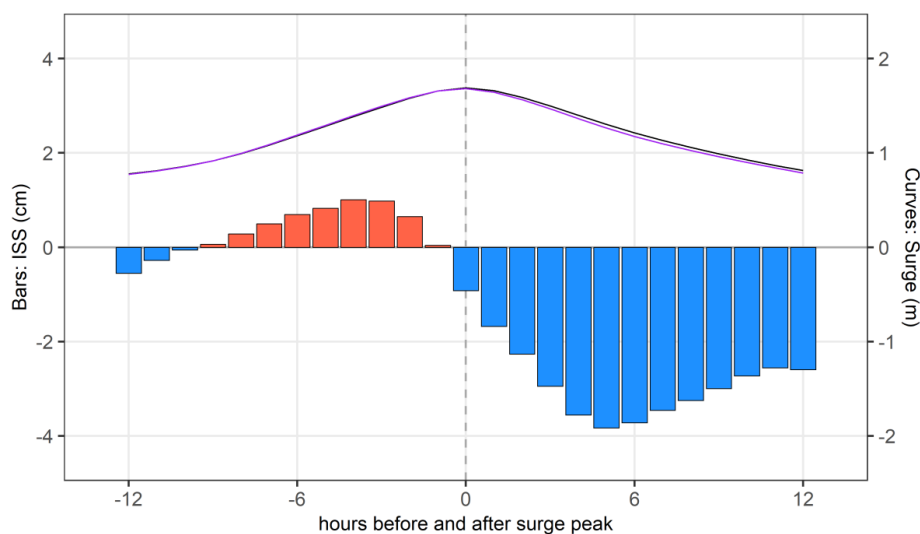
422



423

424 Figure 10. (a) Frequency distribution of the impact of SLR on storm surges (ISS). (b) Map of ISS
 425 for the surge events detected at three locations within a 1 h time window.

426



427

428 Figure 11. Event-averaged ISS (bars, cm), shown relative to the timing of surge peaks (gray
 429 dashed line). The curves (m) represent storm surge (black) and storm surge under SLR minus 1
 430 m (purple).

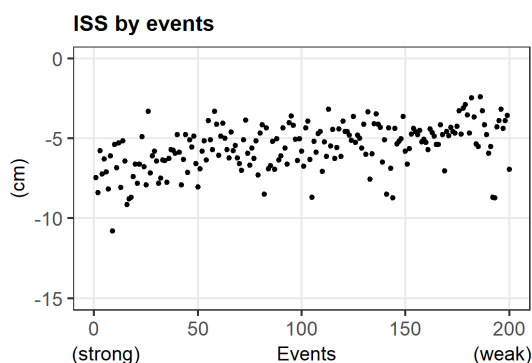
431

432 The reduction is on the order of centimeters for 1 m SLR, comparable to the SLR impact on
 433 TSI at surge peaks (Fig. 4). On average, surge maxima decrease by about 1 %. Most of the
 434 strongest reductions fall between -3 cm and -8 cm (5th-95th percentiles), with a mean of
 435 -5 cm (Fig. 12). Larger reductions occur during stronger surge events, suggesting that
 436 stronger currents enhance the underlying interactions. Overall, these results demonstrate
 437 that TSI is a dominant control on both the timing and magnitude of surge-related water



438 levels. In contrast, the effects of SLR on these processes are comparatively small, although
439 systematic and physically consistent.

440



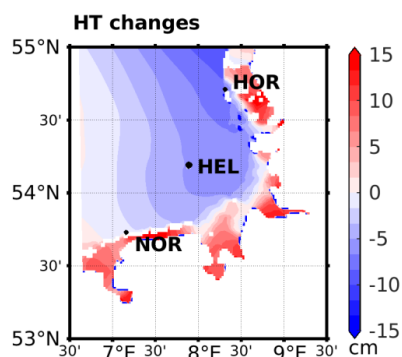
441

442 Figure 12. The maximum reduction of ISS, ordered from highest to lowest surge.

443

444 3.4 Impact of SLR on tides (IST)

445 Changes in tidal dynamics due to SLR were not the main scope of this study. However, to ensure
446 consistency and to set our results into the context of previous studies, an analysis of changes in
447 high tide under SLR was made based on the tide-only simulations with and without SLR
448 described in Section 2.3. For that we compared high tide magnitudes associated with the
449 highest 200 storm surge events analyzed in Section 3. Figure 13 shows a mean difference of
450 these high tides deviating from the SLR magnitude. High tide decreases in response to SLR for
451 the offshore areas of the German Bight, especially in the eastern part and increases for the
452 shallow areas of the Wadden Sea. This pattern generally agrees with the pattern of the M2
453 amplitude changes due to a SLR of 0.8 m shown by *Rasquin et al.* (2020) or SLR of 0.8 m and 2 m
454 shown by *Jordan et al.* (2021).



455



456 Figure 13. Map of high tide changes in response to SLR.

457

458 **4 Discussion**

459 **4.1 High tide changes due to SLR**

460 As can be inferred from previous studies, the response of the M2 tidal amplitude and high
461 water level to SLR is strongly dependent on experimental design. In particular, the
462 representation of coastal geometry plays a key role: simulations that allow coastal flooding
463 produce markedly different spatial patterns compared to those assuming rigid coastlines
464 (e.g. Pelling et al., 2013; Idier et al., 2017), with the present study being closer to the latter
465 configuration. Additionally, the resolution of the underlying bathymetry, and likely the
466 numerical model resolution itself, influence the response on a qualitative level (Rasquin et
467 al., 2020). They have demonstrated that the larger increase in dissipation rate in the fine
468 bathymetry model is largely due to overall increased current speeds, which cannot be seen
469 to the same extent from a coarse bathymetry model. Overall, the response of tides on SLR in
470 the German Bight is collectively determined by the interplay of the reduced dampening of
471 high water due to increased water depth and the consequent weakening of bottom friction,
472 increased dissipation rates due to finer resolution and additional increase in dissipation due
473 to depth reduction when coastal flooding of grid cells is considered.

474 Moreover, the magnitude of the imposed SLR further modulates the spatial pattern of M2
475 amplitude and high tide changes. Under moderate SLR (approximately 0.8–1 m), a substantial
476 part of the German Bight exhibits a reduction in M2 amplitude when estimated with finer
477 bathymetry resolution (Rasquin et al., 2020; Jordan et al., 2021), and an increase in M2
478 amplitude (or high water) when coarser bathymetry is implied (Idier et al., 2017; Rasquin et al.,
479 2020). For higher SLR scenarios (typically 2 m), the area of reduced amplitude diminishes and
480 shifts northward (Pelling et al., 2013; Idier et al., 2017; Jordan et al., 2021). In contrast, extreme
481 SLR scenarios (5–10 m) lead to a consistent positive-valued signal in M2 amplitude (Rasquin et
482 al., 2020) or maximum high water (Idier et al., 2017) across the German Bight independently on
483 the bathymetry resolution. For the moderate SLR scenario considered in the present study, the
484 spatial pattern of changes is consistent with the general consensus of the abovementioned
485 studies concerning the response of M2 amplitude. It should be noted, however, that present
486 study focuses on high tides across the full tidal spectrum rather than directly on M2 amplitude.
487 These metrics are not equivalent, as high tides reflect the combined effect of multiple tidal
488 constituents but obviously omit the low tide changes, which directly affect tidal amplitude.

489



490 4.2 Importance of timing

491 We show statistically that the most pronounced negative TSI values occur mostly near high tide,
492 not at the surge peaks, and the oscillation of TSI is highly dependent on the tidal phases. The
493 average TSI is about -38 cm near high tide, and -21 cm at the surge peaks. Similar values of TSI
494 were reported in previous studies. *Rego and Li (2010)* showed TSI are -30 cm at high tides for
495 the diurnal tide simulations and -40 cm for the semidiurnal cases in their case study. The values
496 are also comparable to the value of -32 cm reported previously in a case study (*Song et al.,*
497 *2020*) and fall within the ranges of another case study of *Huang et al. (2025)*. *Zhang et al. (2017)*
498 focused on the TSI at surge peaks and highlighted the modification of tidal phases on the
499 maximum surges. In their analysis of multiple typhoon events, TSI was found to dampen the
500 surge peak near high tide and to enhance it when occurred during low tide by about 15–25 %.
501 They also underscored that the TSI increases the duration of storm surge while reducing the
502 maximum surge, and vice versa for the low tide cases. Although water level enhancement
503 during low tide may not be critical for EWLs, it can prolong the periods of elevated low water
504 during long-lasting storm surge events that span several tidal cycles. This may pose challenges
505 for the drainage of low-lying coastal areas, as well as increase the risk of compound flooding by
506 limiting discharge capacity (e.g. *Bormann et al., 2024*).

507

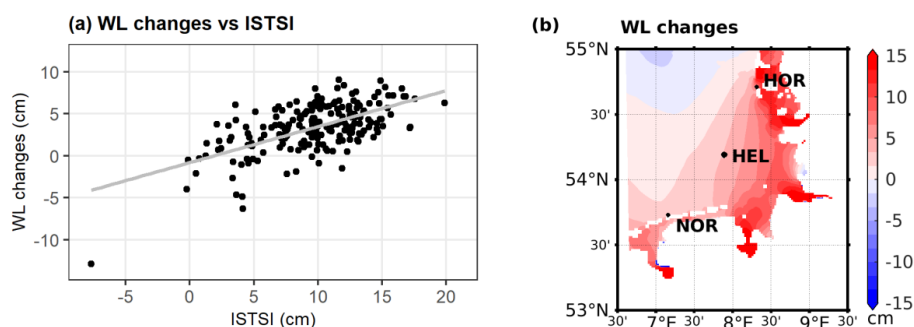
508 4.3 EWL changes, attribution and limitations

509 Taken together, the results highlight that SLR influences EWLs through changes in tides (IST),
510 storm surges (ISS), and their nonlinear interaction (ISTSI), which need to be interpreted jointly
511 to understand the total response. In particular, a weakening of TSI under SLR (Fig. 8 and 9)
512 reduces its dampening effect on water level, contributing to higher extremes. As shown in Fig.
513 14, WL changes due to +1 m SLR are largely concentrated between -1 cm to 7 cm (5th–95th
514 percentiles). The changes are in similar order of magnitude as shown, e.g., by *Idier et al. (2017)*.
515 These changes exhibit a linear dependence on ISTSI, indicating that the increase in water level
516 beyond SLR alone can be attributed to the weakening of the negative TSI effect under SLR.
517 Detected response may, however, differ significantly when larger SLR is considered (5–10 m), as
518 is indicated, e.g., by *Idier et al. (2017)* or *Rasquin et al. (2020)*. We explicitly omit these low
519 probability scenarios (*Fox-Kemper et al., 2021*) in order to concentrate on results and processes
520 more relevant for the coming decades rather than centuries.

521 It is also worth mentioning that changes in EWL are the result of many factors. Here we only
522 discussed the nonlinear changes and effects of TSI related to the increased MSL, while it is not
523 the only parameter changing with time or climate state. One process indirectly linked to the
524 changing MSL and not considered in the present study is the potential change in bathymetry
525 due to natural variability as well as altering of sediment transport patterns under SLR (*Huisman*
526 *et al., 2022*). This may be of particular importance for the EWL changes in shallow areas, and so



527 far the effect is inconclusive, especially considering the large uncertainty regarding the reaction
 528 of intertidal flats to SLR. Another major contributor to the changing EWL is the change in wind
 529 climate and storminess, which may directly trigger changes in storm surge magnitude under the
 530 climate change conditions (Lowe et al., 2001; Gaslikova et al., 2013; Vousdoukas et al., 2016)
 531 but are out of the scope of present study.



532
 533 Figure 14. (a) Scatter plot of WL changes due to SLR versus ISTSI with fitted regression line. (b)
 534 Map of WL changes due to SLR.

535

536 4.4 Nonlinear versus linear

537 Finally, this study also aims to quantify uncertainty in estimating absolute EWLs under SLR
 538 conditions, which arises from the linear consideration of different WL components as
 539 opposed to a dynamical combination that accounts for nonlinear interactions. For that four
 540 configurations of EWL estimate are compared (Fig. 15) ranging from fully dynamic (WL_SLR)
 541 to fully linear (Linear_0). The statistics of the present-day WL linearly combined with 1 m SLR
 542 (WL+SLR) and the dynamically simulated WL under SLR (WL_SLR) are close to each other,
 543 with the mean values underestimated by about 1 % in case of linear combination. Larger
 544 differences emerge under surge conditions (Fig. 15a) when tide, surge and 1 m SLR are all
 545 added linearly (Linear_0), thus omitting surge-tide-SLR interactions altogether, and when
 546 surge-SLR and tide-SLR interactions are considered by using the sum of tide and surge under
 547 SLR (Linear_1). Both approaches exclude TSI processes and overestimate the mean WL_SLR
 548 by about 11 % and 9 %, respectively. Under non-surge conditions (Fig. 15b), the statistics of
 549 the four estimates are very similar to each other, with the largest difference of about 4 %
 550 occurring also between the fully dynamic and fully linear estimates.

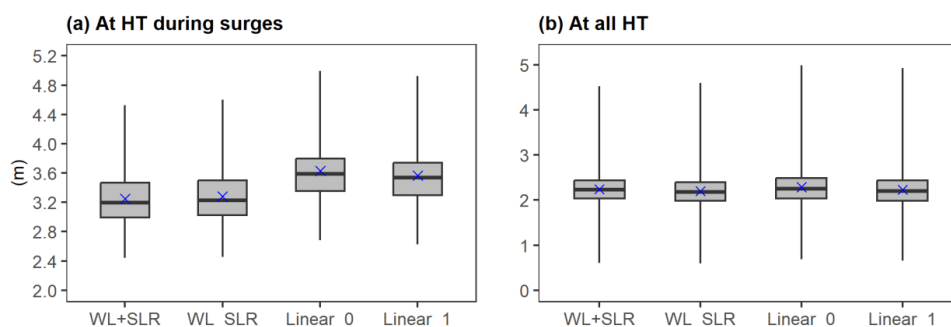
551 These results highlight the importance of TSI as a source of uncertainty in water level
 552 simulations. The nonlinear effects of SLR on oceanic components remain small (1–2 %), even
 553 under surge conditions, corresponding to a few centimeters for 1 m SLR. However, these
 554 effects are systematic and therefore relevant for process understanding and uncertainty



555 quantification. This is consistent with earlier studies showing limited nonlinear influence of
556 SLR on surge propagation (Lowe et al., 2001; Sterl et al., 2009).

557 Overall, nonlinear processes do not invalidate linear superposition approaches but instead
558 define the associated uncertainty. Consequently, the inclusion of nonlinear effects depends
559 on the application: linear addition of SLR is sufficient for many practical purposes, whereas
560 nonlinear processes primarily introduce small but systematic deviations that are relevant for
561 more precise assessments.

562



563

564 Figure 15. Boxplots of water levels (a) at high tides under surge condition and (b) at all high
565 tides: FULL_0 plus 1 m SLR linearly (WL+SLR), FULL_1 from the dynamical run with SLR
566 (WL_SLR), linear sum of TIDE_0, SURGE_0 and 1 m SLR (Linear_0), and linear sum of TIDE_1 and
567 SURGE_1 minus 1 m (Linear_1). The boxplot displays the minimum, 25th percentile, median, 75th
568 percentile and maximum values, and the blue cross marks the mean.

569

570 5 Conclusions

571 This study investigates how sea level rise (SLR) modifies storm surges and tide-surge interaction
572 (TSI) in the German Bight, with the aim of assessing the relevance of nonlinear processes for
573 estimating extreme water levels (EWLs). Using long-term hydrodynamic simulations and
574 sensitivity experiments, we separated tidal, meteorological, and interaction effects and
575 quantified their response to a +1 m SLR scenario.

576 Our analysis shows that the influence of TSI on storm surge height and timing is strongly
577 controlled by tidal phases. The largest dampening occurs when surge peaks coincide with high
578 tide, while the interaction is weaker and occasionally constructive near low tide. In addition, TSI
579 shifts peak surge levels away from high tide, further modifying the temporal structure of
580 extreme events. The strength of TSI increases with surge magnitude, whereas its dependence
581 on tidal characteristics such as high tide level or tidal range is comparatively weak and more
582 variable.



583 Under SLR, the negative effect of TSI is slightly weaker. This reduction in nonlinear interaction is
584 consistent with the water depth increase and the associated decrease in bottom friction, which
585 leads to a more linear system response. The weakening primarily manifests as a reduction of the
586 negative TSI signal by about 10 cm on average, corresponding to a decrease of roughly one
587 quarter of its present-day magnitude. As a consequence, the dampening effect of TSI on surge-
588 related water levels is reduced under SLR. This contributes to an increase in total water level
589 beyond SLR alone, although the effect remains small compared to the direct SLR contribution.

590 Moreover, storm surge heights decrease slightly under SLR when considered as a stand-alone
591 process. This reduction is small (a few centimeters for 1 m SLR) and is attributed to two depth-
592 related mechanisms: a phase shift associated with faster wave propagation in deeper water and
593 the reduced efficiency of wind stress in driving a thicker water column. The magnitude of this
594 effect scales with surge intensity but remains small in absolute terms.

595 In summary, this study demonstrates that TSI contributes largely to the uncertainties in water
596 level simulations, whereas nonlinear SLR effects are small but play a systematic role in shaping
597 the dynamics and associated uncertainties in shallow coastal seas. The findings support the use
598 of linear superposition of SLR and present-day EWL estimates for many applications.

599 Meanwhile, they highlight that nonlinear interactions introduce process-based adjustments,
600 which are particularly relevant for a detailed understanding of coastal dynamics and for
601 quantifying uncertainties in high-precision applications, such as coastal protection design or
602 attribution studies.

603

604 *Data availability.* The model outputs used in this study are available upon request from the
605 authors.

606

607 *Author contributions.* XL performed the analysis. All authors contributed to the study design,
608 analysis discussion and manuscript preparation.

609

610 *Competing interests.* The authors declare that they have no conflict of interest.

611

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613

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617



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