

# [Severe flooding in the atoll nations of tuvalu and kiribati triggered by a distan...](https://assignbuster.com/severe-flooding-in-the-atoll-nations-of-tuvalu-and-kiribati-triggered-by-a-distant-tropical-cyclone-pam/)

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## Introduction

Coastal areas are perceived to be at risk of increasingly frequent and severe flooding and erosion impacts associated with sea level rise (SLR). This has led to concern that island nations are especially vulnerable to coastal flooding ( [Nicholls et al., 2007](#B26) ; [Seneviratne et al., 2012](#B33) ), particularly the atoll nations of Kiribati, Tuvalu, Marshall Islands and Maldives ( [Barnett and Adger, 2003](#B2) ; [Nicholls et al., 2011](#B25) ). There has been some debate on the nature and timing of these impacts, however. Several studies have presented evidence that, despite sea level rise on the order of 20 cm over the last century, atoll islands have not (yet) experienced net erosion and that many may have some capacity to keep up with future SLR (e. g., [Webb and Kench, 2010](#B44) ; [Beetham et al., 2017](#B4) ; [Tuck et al., 2019](#B38) ) through the deposition of reef-derived materials during storm over wash events (e. g., [Maragos et al., 1973](#B22) ; [Smithers and Hoeke, 2014](#B34) ). Other studies have suggested that rates of SLR will outstrip reefs accretion rates ( [Perry et al., 2018](#B29) ), and that in any case, the increased frequency and severity of inundation will make most atoll islands uninhabitable within a few decades (e. g., [Storlazzi et al., 2018](#B36) ). Regardless, there is widespread agreement that overall vulnerability to SLR will be heterogeneous due to differences in reef/island morphology and exposure to extreme sea level drivers (viz sea level variability, storm waves, storm surges, and astronomic tides), and that more detailed event-based case studies are needed to understand the role of background sea level (and SLR) compared to these drivers ( [Woodroffe, 2008](#B45) ; [Beetham and Kench, 2018](#B3) ; [Idier et al., 2019](#B19) ). Such case studies can also better inform the development of early warning forecast systems, which are essential in the face of likely increased flooding and erosion events ( [Storlazzi, 2018](#B35) ), by identifying the relative importance of the extreme sea level processes and indications of local vulnerability.

In this study, we investigate extensive flooding and erosion impacts which occurred in Tuvalu and the Gilbert Islands of Kiribati during March 2015. Unlike the case studies of [Hoeke et al. (2013)](#B17) , [Wadey et al. (2017)](#B42) , [Ford et al. (2018)](#B15) , and [Wandres et al. (2020)](#B43) , all of which focused at least primarily on events triggered by distance-source swell waves generated by mid-latitude (temperate) cyclones in the western Pacific (primarily the Federated States of Micronesia, Papua New Guinea, and Solomon Islands), the Maldives, Majuro Atoll (Marshall Islands) and Fiji, respectively, the events in Tuvalu and Kiribati described here are associated with the passage of tropical cyclone (TC) Pam. TC Pam formed in the central south Pacific in early March 2015. It reached a category five severity and made landfall or otherwise directly impacted several islands in Vanuatu, causing widespread damage and loss of life (Gov. of [Vanuatu, 2015](#B40) ). It then moved along a southerly track between Fiji and New Caledonia, exiting the region around March 15th. The resulting swell propagated throughout the central Pacific, causing flooding and damage to communities in Tuvalu, Kiribati, and Wallis and Futuna, all over 1, 000 km from TC Pam’s track ( [Damlamian et al., 2017](#B9) ; [Figure 1](#F1) ).

FIGURE 1

Tropical Cyclone (TC) Pam’s track (black line), with maximum hindcast significant wave height (Hs) during the TC’s passage (March 6–17, 2015) indicated in color shading. Island nations/archipelagos in the area are labeled and their exclusive economic zones ( [Flanders Marine Institute, 2019](#B14) ), indicated with gray lines. The locations in [Table 1](#T1) are indicated with black +’s.

Here, we focus on the remote impacts to Tuvalu and Kiribati, which were poorly forecast or otherwise anticipated ( [Taupo and Noy, 2017](#B37) ). Multiple reports were collated ( [Table 1](#T1) ) and then a combination of astronomical tidal predictions and numerical wind-wave and sea level hindcasts, spanning from 1979 to the present, were combined using a total water level (TWL) approach (e. g., similar to [Dodet et al., 2019](#B10) ), to estimate the climatological conditions and factors contributing to recorded impacts. The accuracy of these TWL components were independently assessed using tide gauge and satellite altimetry observations; it also includes an empirical wave runup estimate. This resulting proxy for flooding severity was compared to the recorded flooding and erosion impacts; the relative contribution of waves, tides and background sea level to TWL were assessed, as was event duration and wave direction.

TABLE 1

Documented remote impacts in Tuvalu and Kiribati.

## Materials and Methods

### Summary of Impacts to Islands and Atolls

Information on which island and atolls within Tuvalu and Kiribati were impacted (or were not), and types of impacts were collated from a number of sources, primarily disaster relief agencies. These are summarized in [Table 1](#T1) ; these locations are also plotted in [Figure 2](#F2) . Note that (hand drawn) “ inundation” and “ erosion” maps, and notes provided by the Tuvalu Public Works Department to the authors are provided in this article’s [“ Supplementary Material”](#S9) section.

FIGURE 2

Region and locations indicated in [Table 1](#T1) . The maximum hindcast significant wave height ( *H* x ) during TC Pam’s passage (March 6–17, 2015) is shown in subplot(a), which is normalized by the local 1-in-20 year annual exceedance probability (AEP) shown in subplot(b). Note *H* s , rather breaking wave height ( *H* b ) or total water level ( *TWL* ) as these latter quantities follow wave shoaling and breaking across shallow fringing reef, respectively, and the area of [Figure 2](#F2) (apart from the scattered islands and reefs) contains abyssal ocean depths.

### Geophysical Data

To assess the timing and location of regional TCs, particularly TC Pam, the IBTrACS ( [Knapp et al., 2010](#B21) ) database (version 04r0) [1](#footnote1) was interrogated for TCs tracks; any portion of which occurred between longitudes 155°E and 170°W and latitudes of 30°S and 10°N and between the years 1979–2019 (the time period of the wave hindcast used in this study, see next section) were retained and defined as “ regional” TCs. TC Pam’s track is given in [Figure 1](#F1) .

To assess impacts, three different categories of sea level information were used in this analysis: wind-waves, astronomical tides, and non-tidal sea-level. These are detailed in following sub-sections.

#### Wind-Waves

Hourly significant wave height ( *H* s ), integrated wave energy flux ( *C g E* ), peak wave period ( *T* p ) and peak wave direction ( *D* p ) for years 1979–2019 were extracted from the CAWCR Wave Hindcast, a validated global wind-wave hindcast with increased (approximately 7 km) spatial resolution around the Pacific island nations ( [Durrant et al., 2014](#B11) ) [2](#footnote2) and related collections. Previous global and regional comparisons with satellite altimetry and wave buoy data have found the overall hindcast *H* s bias and overall root-mean-square (RMS) difference to be on the order of −5% and < 10%, respectively, in the central Pacific ( [Durrant et al., 2014](#B11) ; [Hemer et al., 2016](#B16) ). However, in this study, a further verification of the hindcast’s ability to estimate the extreme waves associated specifically with TC Pam was performed by comparing hindcast *H* s to calibrated *H* s measured by multiple satellite altimeters, provided via the [Ribal and Young (2019)](#B30) dataset (sourced from Australia’s Integrated Marine Observing System) [3](#footnote3) . While overall correlation coefficients ( *R* ) and errors between hindcast *H* s and that estimated by various altimeters/bands indicate good agreement ( *R* > 0. 9, std. err. < 0. 004 m) for the area and time period indicated in [Figure 1](#F1) , the hindcast overestimated the most extreme waves ( *H* s > 12) near TC Pam’s eye wall compared to the altimeter data by up to several meters ( [Supplementary Figure 1](#S9) ). This divergence is at least partially attributable to (1) the known tendency of the WaveWatch III model to overestimate energy of waves traveling in the oblique and opposing winds (as would occur in a tight-radius TC, including with the “ ST4” physics package used in the hindcast) and; (2) uncertainty in the radar altimeters estimation of *H* s in such extreme conditions: [Ribal and Young (2019)](#B30) limit their calibrations to values of *H s* < 9. 0 m. This overestimation (or divergence between hindcast and altimetry observations) of *H* s does not occur at all for the same time period in the area indicated by [Figure 2](#F2) (which corresponds to the area of flooding and erosion impacts researched in this paper), and statistics in this area are modestly improved ( *R* > 0. 92, std. err. < 0. 002 m, [Supplementary Figure 2](#S9) ). For further information on the details and discussion of this hindcast/satellite altimeter comparison, please refer to [Supplementary Material](#S9) , however, the authors have confidence that the CAWCR wave hindcast represented the propagation of the wave field associated with TC Pam, and its representation within the overall extreme value statistics of the hindcast, throughout the Gilbert Islands and Tuvalu sufficiently to support the paper’s conclusions.

#### Astronomical Tides

Hourly astronomic tidal predictions ( *η tide* ) were calculated for a time period corresponding to that of the wave hindcast (years 1979–2019) from TPXO tidal constituents (version 9. 1, [Egbert and Erofeeva, 2002](#B13) ) at all locations in [Table 1](#T1) . To assess the skill of TPXO-based tidal predictions used in this study, they were compared to tidal predictions based on hourly tide gauge observations from Tarawa and Funafuti (years 1988–2018 and 1979–2017, respectively) sourced from the University of Hawaii Sea Level Center ( [4](#footnote4) [Caldwell et al., 2015](#B5) ). Root-mean-square-differences (RMSD) between tide gauge and TPXO predictions at Funafuti and Tarawa were 0. 01 and 0. 02 m, respectively. More importantly for this analysis, 99. 5th quantiles differences between tide gauge and TPXO predictions at Funafuti and Tarawa were 0. 021 and 0. 025 m, respectively. The tidal software UTide ( [Codiga, 2011](#B8) ) was used both to estimate constituents from tide gauge observations and to calculate predictions for both the TPXO and tide gauge constituents.

#### Non-tidal Sea-Level

The computation of TWLs and associated extreme value analysis (see next) requires an uninterrupted time span of non-tidal sea-level data that corresponds to the wave hindcast (years 1979–2019) and the astronomical tides calculated in section “ Astronomical Tides”. This precludes using sea surface heights measured by satellite altimetry, which only start around 1992. We evaluated three global gridded sea level products, all available from publicly accessible data repositories, which meet this time span requirement:

1. NCEP Climate Forecast System Reanalysis Reforecast (CFSR) 6-hourly sea surface height ( [Saha et al., 2013](#B32) );

2. CSIRO monthly sea-level reconstruction ( [Church and White, 2011](#B7) ); and

3. ORAS5 Ocean ReAnalysis System 5 (ORAS5) monthly sea surface height [5](#footnote5) .

These sea level products were assessed for best representing background non-tidal sea levels ( *η SL* ) by comparing them to the 30-days median low-pass filtered tide gauge residual sea level at Tarawa and Funafuti (see section “ Astronomical Tides”). All sea levels, (including tide gauge residuals) where adjusted such that mean sea level = 0 between 1999 and 2009 prior to comparison. Correlation coefficients (R) and RMSD calculated against the low-pass filtered tide gauge residuals are shown in [Table 2](#T2) .

TABLE 2

Comparison of CFSR, CSIRO and ORAS5 sea surface height values with monthly-median low-pass filtered non-tidal (residual) tide gauge observations.

ORAS5 clearly outperforms the other two sea surface height products in comparison to the (observed) Tarawa and Funafuti tide gauge residuals. ORAS5 monthly values are linearly interpolated to the (hourly) times of the wave hindcast (years 1979–2019) to represent non-tidal sea levels ( *η SL* ); throughout the remainder of this article, *η SL* refers to that derived from ORAS5 sea surface height values. It should be noted that thus *η SL* omits high-frequency (timescales < 1 month) sea level variability. However, the variance of the 30-days median low-pass filtered (observed) residuals is 0. 005 m at Tarawa and 0. 010 m at Funafuti. The high-frequency (remainder) residual variance at both sites is 0. 001 m, i. e., 5–10 times less than that of the monthly variance; which is fairly typical of low-latitude Pacific locations where sea level variability is dominated by seasonal and interannual (e. g., ENSO) dynamics ( [Merrifield et al., 1999](#B24) ; [Church et al., 2006](#B6) ; [Zhang and Church, 2012](#B46) ). Thus, the lack of inclusion of higher-frequency sea level fluctuations is assumed to introduce a relatively small error into this analysis. Future studies may evaluate introducing inverse barometer estimates, based on global sea level products to further ameliorate this error.

### Extreme Value Analysis and Total Water Level (TWL)

To reduce complexity of the extreme value (statistical) analysis, we focus on four of the previously defined variables: *C g E* (wave energy flux), *D* p (peak wave direction), *η tide* , and *η SL* as well as event duration, which arises from the analysis. *C g E* , rather than *H* s , is chosen as a primary diagnostic variable for the analysis, since (as a measure of power, rather than energy), it tends to scale more linearly with nearshore processes such as wave runup and setup.

Extreme events are identified for each site in [Table 1](#T1) by choosing a threshold and identifying exceedances of *C g E* above it; an event begins when the *C g E* exceeds the threshold, and a new event is assumed only when *C g E* remains below the threshold for more than 12 h (i. e., “ declustering” events). The threshold for this analysis is the 99. 5% quantile.

Event duration is then calculated as the amount of time that each declustered *C g E* event remains above the threshold. The *D* p (peak wave direction) associated with the event is taken as that occurring at the time of the peak (maximum) *C g E* . The *η SL* and *η tide* associated with the event are the respective maxima of each occurring during the storm duration, i. e., not necessarily the water level at the time of maximum *C g E* . To further aid in interpretation, empirical annual exceedance probability (AEP), also commonly called return periods or return levels, were empirically calculated from the declustered storm event maximum *C g E* values at both locations.

In order to better understand how these variables ( *C g E* , *D* p , *η SL* , and *η tide* ) combine to create extreme sea level events, we follow a “ total water level” (TWL) approach ( [Dodet et al., 2019](#B10) ), and calculate a further variable ( *η 2 )* , defined as the 2% exceedance water level at the shoreline above the “ still” water level, as a function of wave setup and variable swash (“ runup”) due to both wind-waves and infra-gravity/longwave dynamics associated with shoaling and breaking of wave groups at the reef edge. The TWL is simply calculated as:

TWL = ⁢ η SL + ⁢ η tide ⁢ + ⁢ η 2 ( 1 )

To calculate *η 2* , we follow the simpler of the two approaches described by [Merrifield et al. (2014)](#B23) , viz:

η 2 = b 1 ⁢ H b + b 0 ( 2 )

Where *H* b is “ breaking wave height,” calculated using a simplified relation to account for wave refraction and shoaling to breaking depth from the (deep water) *H* s provided by the wave hindcasts:

H b = [ H s 2 ⁢ T p ⁢ ( 4 ⁢ π ) - 1 - cos ⁢ ( D p - D N ) ⁢ γ ⁢ g ] 2 / 5 ( 3 )

In Equation 2, the two empirical coefficients *b* 1 and *b* 0 are assumed a value of 0. 3 and –0. 1, both near the mean value and within the 95% confidence limits for the two study sites examined in [Merrifield et al. (2014)](#B23) . While these sites were both in the Marshall Islands, it is arguable that the Tuvalu and Kiribati sites in this study share similar atoll/narrow fringing reef morphology to the Marshall Islands sites, although morphological effects are discussed in more detail in the following sections. In Equation 3, γ is the ratio of breaking *H* b to breaking water depth and *D* n is the local shore-normal angle.

## Results

Time series of total water level ( *TWL* ) and breaking wave height ( *H* b ) centered on a 2-week period of the peak of *H* b during the time period of TC Pam’s passage, is presented in [Figure 3](#F3) . From these plots, it is clear that most impacted locations experienced a sustained period of 5 or more days of *H* b above the 99. 5th quantile threshold (based on the entire 40-year timeseries); *TWL* was similarly elevated during successive high tides during the same period. These highly elevated values (above the threshold) are noticeably absent for Funafuti, where no impacts were reported during the period; however, Nukufetau and Niulakita also did not report impacts, but experienced highly elevated *TWL* and *H* b , similar to highly impacted neighboring Tuvaluan islands/atolls ( [Table 1](#T1) and [Figure 3](#F3) ). Potential reasons for this discrepancy are discussed later.

FIGURE 3

Time series plot of estimated breaking wave height ( *H b , red* ), wave setup/runup ( *η 2 , black* ) and total water level ( *TWL* , blue) for selected islands listed in [Table 1](#T1) , from March 6—March 20, 2015. The 99. 5th quantiles (percentiles) for the entire hindcast timeseries (1980–2019) are shown as dashed red and blue lines for *H* b and *TWL* , respectively.

As previously mentioned, the timing of high tides in relation to the timing of maximum *H* b (and *C g E* ) appear to have modulated the timing of impacts at the different islands. However, astronomical tides were not abnormally high (mostly roughly between spring and neap, [Figure 3](#F3) ) during the event at any of the impacted islands. This tidal modulation could explain the lesser impacts reported in the southern islands of Tuvalu (except Nukulaelae) where the peak of *H* b (and *C g E* ) was of shorter duration and occurred predominantly during neap tides.

To better examine the relative contribution of *C g E* , tides ( *η tide* ), background sea level ( *η SL* ), other multivariate aspects of this event and its relation to other *TWL* extreme events, [Figures 4](#F4) , [5](#F5) present “ scatterplot matrix” diagrams for Tamana and Nui, which were selected as representative of impacted locations within Kiribati and Tuvalu, respectively. These scatterplot matrices allow the visualization of the co-occurrence *C g E* , *η tide* , and *η SL* , as well as wave direction *(D p )* and event duration for all events above the (99. 5th quantile) *TWL* threshold within the hindcast record, particularly when combined with annual exceedance probability (AEP) plot of all locations ( [Figure 6](#F6) ).

FIGURE 4

Scatterplot matrix of Tamana (Kiribati) event-based data identified by declustering extreme sea level events with integrated wave energy flux ( *C g E* ) above the 99. 5 percentile threshold. The variables are: within-event maximum *C g E* , peak wave direction ( *D* p ) during the hour of maximum *C g E* , within-event maximum astronomic tide maximum ( *η tide* ), background sea level ( *η SL* ), and event duration (hours). Within each subplot the event closest to March 15, 2015 (i. e., associated with TC Pam) is highlighted with a red circle.

FIGURE 5

Scatterplot matrix of Nui (Tuvalu) event-based data identified by declustering extreme sea level events with wave energy flux ( *C g E* ), above the 99. 5 percentile threshold. The variables are the same as in [Figure 4](#F4) . Within each subplot the event closest to March 15, 2015 (i. e., associated with TC Pam) is highlighted with a red circle.

FIGURE 6

Annual exceedance probability (AEP) plots for selected islands listed in [Table 1](#T1) . Black and blue dots indicate empirical AEPs for estimated breaking wave height ( *H* b ), wave runup ( *η 2* ) and total water level ( *TWL* ), respectively; black and blue lines indicate a maximum likelihood estimate curve using a generalized Pareto tail model for the same variables, respectively. 95% confidence limits are indicated with dotted lines. Peak wave direction ( *D* p ) for each event is indicated by red arrow on the *H* b AEPs, respectively. Where *TWL* or *H* b empirical AEP events have occurred during or up to 5 days after the occurrence of a TC in the region (see [Figure 6](#F6) ) are highlighted with green circles and labeled with the TC name; the event associated with TC Pam is highlighted with a red circle. For events with RI > 10 years not matched with a TC are labeled with their date.

At Tamana (Kiribati), maximum *C g E* (and thus estimated wave setup/runup) was by far the highest in the entire hindcast record, on the order of 40% higher than the next highest event in terms of *C g E* ( [Figure 4](#F4) ). The event also had the longest or second longest duration at the impacted islands. This probably contributed to the noted erosion at Tarawa, in addition to recorded severity of inundation at the impacted islands in Kiribati. At Nui (Tuvalu), maximum *C g E* was not the highest in the entire hindcast record, it was third highest ( [Figure 5](#F5) ). However, when combined with *η SL* and *η tide* it created the highest *TWL* in the record ( [Figure 5](#F5) ). Event duration at Nui ( [Figure 5](#F5) ) was also among the top three events, and likely contributed to the extensive erosion recorded at many of the locations in Tuvalu, as in Kiribati.

Overall, *TWL and H b* were the highest in the hindcast period, with a calculated empirical AEP of 40 years (the highest possible for the time duration of the hindcasts used in this study) for all islands in Kiribati with recorded impacts ( [Table 1](#T1) and [Figure 6](#F6) ). These results are more varied in Tuvalu: *H* b empirical *AEPs* generally ranged between 10 and 20 years, in the northern islands, elevated *η tide* and particularly *η SL* increase *TWL* empirical AEP values above those of *H* b . In the southern islands, the aforementioned later onset of elevated *H* b , closer to neap tides, lead to lower *TWL* empirical AEP values relative to that of *H* b . Never the less, all islands in Tuvalu which recorded severe impacts experienced greater than a 10-year AEP *TWL* , excepting Nukulaelae ( [Table 1](#T1) and [Figure 6](#F6) ).

In addition to the *TWL* and *H* b event AEP levels of indicated in [Figure 6](#F6) , wave direction ( *D* p ) associated with each event is also indicated by an arrow indicating the direction waves are traveling at the time of peak *H* b . This indicates that all large events (AEP > 10 years) are from the west in Kiribati, but not in Tuvalu. At some (but not all) Tuvalu locations, TC Pam was the largest westerly event. To further investigate the drivers the storm wave events, all regional TCs with maximum sustained winds greater than 48 knots (the lower threshold for category 2 TCs according to the southern hemisphere intensity scale) were queried. *H* b and *TWL* events which occurred during or up to 5 days after TCs which occurred within ± 5 days of empirical AEP events are labeled as such in [Figure 6](#F6) and these TCs tracks are plotted in [Figure 7](#F7) . This illustrates that despite TC tracks never crossing the Gilbert Islands and only very rarely crossing the Tuvalu islands within the 40-year hindcast period, all *H* b events (and most *TWL* events) greater than the 10-year AEP value at all locations excepting Tarawa (the northernmost of the study site), are associated with TCs. All of the associated TCs’ tracks lay primarily well to the south and west (or north and west in one case) of the Gilbert Islands and Tuvalu, and most, though not all, attained at least a TC category 4 maximum sustained wind intensity. Of note is TC Kina which occurred in early January, 1993. It is associated with the largest *H* b and *TWL* at many of the southern Tuvalu islands in our analysis; severe flooding and destruction of homes were reported on Nanumea, Nanumaga, Niutao, Nui, and Vaitupu [6](#footnote6) .

FIGURE 7

TC tracks in region of the Gilbert Islands (Kiribati) and Tuvalu (1979–2019). TCs which were matched with *TWL* and/or *H* b events above the 99. 5th quantile threshold (i. e., those plotted in [Figures 4](#F4) , [5](#F5) ) are plotted in color, indicating along-track maximum sustained winds; other cyclones are plotted in gray.

Finally, it should be reiterated that at all locations, background sea levels ( *η SL* ) were the highest or among the highest on record (+0. 1 m) during the TC Pam events. This definitely contributed to the overall severity of the event, particularly in Tuvalu. What portion of that is attributable to sea level rise (SLR) vs. seasonal and/or inter-annual (e. g., ENSO) sea level variability is beyond the scope of this paper, since higher order techniques (such as multivariate regression or empirical function analyses) are needed to separate these two components ( [Zhang and Church, 2012](#B46) ; [Albrecht et al., 2019](#B1) ). This will be addressed in future studies.

## Discussion and Conclusion

In this study, we link a widespread series flooding, island overwash and erosion events in 2015 in the atoll nations of Tuvalu and Kiribati to locally extreme values of total water level ( *TWL* ), composed of regional sea-level, astronomic tide and wind-wave components. Similar to other studies (e. g., [Hoeke et al., 2013](#B17) ; [Wadey et al., 2017](#B42) ; [Ford et al., 2018](#B15) ; [Wandres et al., 2020](#B43) ), extreme wind-waves associated with a distant storm (swell) was the proximate cause, or trigger, of the flooding and erosion reported here. In contrast to (or perhaps complimenting) these other studies, the swell triggering this event was generated not by distant mid-latitude storms, but by a distant tropical cyclone (TC Pam). Like these other studies, however, local TWL extremes (and impacts) were highly modulated by the timing of swell arrival relative to local tidal phase ( [Table 1](#T1) and [Figure 3](#F3) ) and by regional sea level anomalies. Other than the swell generated by it, meteorological phenomena directly associated with TC Pam, e. g., local wind setup and inverse barometer effect, appear to have played little, if any, role.

The relative severity of the flooding and erosion impacts amongst the islands and atolls included does not appear to be simply linked to the local values of *TWL* or breaking wave heights ( *H* b ): e. g., maximum *TWLs* and *H* b heights were significantly smaller (approximately 1. 5 m and 30 cm smaller, respectively) in the southern Gilbert Islands (Kiribati) than in Tuvalu ( [Table 1](#T1) and [Figure 6](#F6) ), but reported impacts in the affected Gilbert Islands appear to be nearly as bad, perhaps worse in some cases. Local annual exceedance probabilities (AEPs), however, align much better with the reported impacts; e. g., significant wave heights ( *H* s ), *TWL* and *H* b AEPs in the southern Gilberts were almost universally higher (despite scalar height values being lower) compared to those in Tuvalu ( [Figures 2](#F2) , [6](#F6) and [Table 1](#T1) ). The exceptional duration of the event (greater again in the Gilberts than in Tuvalu, [Figures 4](#F4) , [5](#F5) ) undoubtedly also played a role. In other words, when extreme water levels associated with TC Pam are taken into local historical context (through extreme value analysis), they align better with the reported impacts. This makes sense from a geomorphology perspective, since local storm ridges and other reef island features are highly dependent on local wave climate (e. g., [Woodroffe, 2008](#B45) ; [Vila-Concejo and Kench, 2017](#B41) ). It also makes sense from a coastal engineering perspective (coastal defenses are typically designed using wave and/or water level exceedance probabilities); and also with respect to traditional island settlement patterns, although inconsistent coastal management and contemporary changes in settlement patterns may cloud this picture somewhat ( [Duvat, 2013](#B12) ; [Kench et al., 2018](#B20) ).

There is also further reason to use such an approach, i. e., to describe (or predict) an island flooding and/or erosion event in the context of an extreme value analysis (EVA) of historical baseline data. In many cases, the EVA can establish a proxy or threshold for anticipated impacts, viz rather than simply predicting flood levels of so-and-so many meters, they can be predicted in terms of a known historical event or (e. g.) the 1-in-50 year level. For example: we predict the flooding impacts of TC Pam will be similar to/less than/greater than (depending on island) TC Kina in 1993 (which did indeed cause similar impacts in many of the southern Tuvalu islands, see “ Results” section). This offers an opportunity to timely, easily understood information toward an increased efficiency of post disaster emergency efforts, and also provide a pathway for the development of a simple regional coastal inundation forecasting (or warning) system for locations where heights and datums of local coastal defenses may not be well know, and with computational need tailored to the current local resources.

Both applications would require further investigation to determine the reliability of existing global or regional wave forecast products to adequately predict TC- or mid-latitude storm generated waves and their subsequent propagation thousands of kilometers away: the wave hindcast used (and verified for TC Pam) in this study is approximately an order of magnitude higher spatial resolution than wave products available from the major numerical weather prediction centers for the region of this study. Ongoing efforts by regional organizations to provide downscaling of these global wave forecast products for Kiribati and Tuvalu could significantly contribute to the generation of island community scale actionable warning information; however, bias correction between forecast products and hindcast/historical products is necessary, and better overall skill assessment of both are needed. To this end, an increase in *in situ* wave observations and monitoring is required, perhaps to at least the level currently provided for sea level monitoring via the tide gauge network. Also required is improved consistency and detail in the reporting of flooding and erosion impacts; despite gathering reports from multiple sources in this study, local impacts at many islands is unclear because of inconsistencies in reporting ( [Table 1](#T1) ). Establishment and widespread adaptation of reef island flooding reporting guidelines would be extremely helpful in this regard.

While this *TWL* extreme value analysis approach appears to have been largely successful at describing the severity of this event at island scale, a number of issues are evident. For example, severe flooding was at Nukulaelae (Tuvalu), while TWL empirical AEP level was amongst the lowest of the impacted sites; conversely Niulakita (Tuvalu) which recorded no impacts had a (much higher) event TWL AEP of ∼10 years. Such discrepancies may of course be attributable to inaccuracies associated with the wave and/or sea level hindcasts. More likely, however, it is due to the highly simplistic approach taken in estimating wave runup and setup ( *η 2* ), which neglects any details of local reef morphology, as well as (other) natural or man-made coastal defenses or shoreline orientation (although the method is capable of the latter). It thus neglects any approximations of local wave transformations required to predict sub-island scale extreme water levels and subsequent flooding. A number of researchers are developing meta-modeling methods (e. g., [Pearson et al., 2017](#B28) ; [Beetham and Kench, 2018](#B3) ; [Rueda et al., 2019](#B31) ) capable of including local details of reef morphology; a number of statistical/analytic methods based purely on a limited number of nearshore observations ( [Merrifield et al., 2014](#B23) ; [Wandres et al., 2020](#B43) ) are also under development. Both of these approaches show promise for much more accurate estimation of local wave transformations over reefs (as relevant to island flooding) compared to the extremely simple approach taken in this study. They all also come with a low enough computational cost to be widely applied to the approach presented in this paper, e. g., by using the kind of long-term hindcast data presented here as boundary condition to establish much more locally relevant extreme value analysis-thresholding. However, these meta-model and statistical nearshore wave transformation models require detailed spatial data on reef morphology (reef flat widths, shoreline orientations, and slopes, etc.) and/or local *in situ* observations of water levels and waves, which are currently available only in very few locations. Also, further research into how to incorporate reef passes and lagoons into such hybrid (statistical-dynamical) approaches is needed.

In summary, the type of TWL event-driven extreme value analysis presented in this paper shows promise to be a valuable tool toward improved understanding and prediction of inundation and erosion events, at both weather forecasting and for longer term future climate timescales. Improved understanding through development of such tools is generally seen as imperative to understand the continued habitability of atoll islands in the face of sea level rise ( [Kench et al., 2018](#B20) ; [Storlazzi, 2018](#B35) ). Although it is beyond the scope of this paper to quantitatively attribute what role sea level rise may have played in the events reported here, it does implicate positive sea level anomaly, and forms a basis to examine attribution of sea level rise in reef island flooding events in future studies.

## Data Availability Statement

The datasets used in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found in the article/ [Supplementary Material](#S9) .

## Author Contributions

RKH conceived the study and preformed most of the analysis and writing. HD, JA, and MW contributed data and other information, as well as assisting with writing and analysis. All authors contributed to the article and approved the submitted version.

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## Conflict of Interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

The handling editor declared a past co-authorship with several of the authors RKH, JA, and MW.

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## Supplementary Material

The Supplementary Material for this article can be found online at: https://www. frontiersin. org/articles/10. 3389/fmars. 2020. 539646/full#supplementary-material

## Footnotes

1. [^](#footnote1a) [https://www. ncdc. noaa. gov/ibtracs/](https://www.ncdc.noaa.gov/ibtracs/)
2. [^](#footnote2a) [https://data. csiro. au/collections/collection/CIcsiro: 6616v8](https://data.csiro.au/collections/collection/CIcsiro: 6616v8)
3. [^](#footnote3a) [http://imos. org. au/](http://imos.org.au/)
4. [^](#footnote4a) [https://uhslc. soest. hawaii. edu/](https://uhslc.soest.hawaii.edu/)
5. [^](#footnote5a) [https://www. ecmwf. int/en/research/climate-reanalysis/ocean-reanalysis](https://www.ecmwf.int/en/research/climate-reanalysis/ocean-reanalysis)
6. [^](#footnote6a) [https://reliefweb. int/report/tuvalu/tuvalu-cyclone-nina-jan-1993-dha-undro-situation-report-1](https://reliefweb.int/report/tuvalu/tuvalu-cyclone-nina-jan-1993-dha-undro-situation-report-1)

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